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**A Quantitative Analysis of the Production, Selection, and  
Career Paths of Texas Public School Administrators**

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**A Quantitative Analysis of the Production, Selection, and  
Career Paths of Texas Public School Administrators**

by

**Bradley Walter Davis B.S.; M.S.**

**Dissertation**

Presented to the Faculty of the Graduate School of

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### **Dedication**

This dissertation is dedicated to Walter D McCracken, my maternal grandfather. He was a World War II veteran, an insatiable consumer of research, and for those that were privileged enough to know him - one of the most intelligent persons they could ever hope to encounter. He was self-educated.



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*this accomplishment would not have been possible without  
your support. I love you more than anything and eagerly  
look forward to the rest of our lives together.*

**A Quantitative Analysis of the Production, Selection, and  
Career Paths of Texas Public School Administrators**

Bradley Walter Davis, Ph.D.

The University of Texas at Austin, 2012

Supervisor: Mark A. Gooden

Using state-wide, longitudinal data on Texas public school educators employed between the 1991-1991 and 2010-2011 school years, this study explores the disproportionate selection of campus leaders based on ethnicity and gender. Through a combination of descriptive and inferential techniques, this study illustrates how trends in the production, selection, and career paths of administratively-certified educators at the various intersections of ethnicity and gender have changed over time. Controlling for a variety of individual work history and campus characteristics, this study also explores how an administratively-certified educator's ethnicity and gender affect their probability of procuring a campus leadership position.

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## **Chapter One: Introduction**

As the 21<sup>st</sup> century grew near, many scholars of educational administration were focused on the purported, upcoming, nation-wide shortage of certified school administrators. In the years that followed, studies began mounting that illustrated this shortage was largely overstated. Since that time, two new research foci have emerged in the field; 1) measuring and improving the effectiveness of leadership preparation programs (LPPs), and 2) understanding the production, distribution, and career paths of school administrators. This study contributes to the knowledge of the latter pursuit.

Over the last decade, educational administration research has expanded our knowledge of the leadership pipeline. We are learning more about the candidates entering LPPs and seeking administrative positions in our public schools. For the first time we have large-scale, quantitative research providing information on the distribution and career paths of campus administrators in certain states. Still, much is to be learned about the administrator selection process for the greater part of the country. Further still, much is to be learned about the

conditions contributing to administrators' moves in and out of schools and the education system itself.

While the aforementioned studies provide nuanced information about administrator distribution and career paths in certain parts of the nation, their collective generalizability cannot account for the distinct and varying contexts within the state of Texas. Nor do they deeply explore the social factors at play in the sorting of educators into and amongst administrative positions.

Naïve analysis of state-wide administrative data on Texas public school educators reveals striking disparities in administrative role-assignment across both ethnicity and gender. Certain ethnicity groups have a much larger proportion of administratively-certified individuals in campus leadership positions than others. Even when assuming assignment to campus administrative position, certain ethnicities are more likely to be in the head principal position than the assistant principalship. Similar disparities exist across gender boundaries. Experiences of different groups become especially compelling when the various intersections of ethnicity and gender are considered.

## Problem Statement

Table 1.1 contains counts of all Texas public school administrators in campus leadership positions during the 2010-11 school year, disaggregated by ethnicity and role (whether an assistant or head principal). Table 1.1 also contains, in the right-hand column, a ratio representing the proportion of individuals within each ethnicity that are assistant principals to the number that are head principals.

**Table 1.1 - Asst. prin to Prin ratio by ethnicity**

	Asst. prin	Prin	AP to Prin ratio
Black	1515	833	1.82
Latino	2345	1599	1.47
White	5302	5117	1.04

The ratios in table 1.1 illustrate clear boundaries, constructed around ethnicity, in the personnel assignments of Texas public school administrators. Whites are equally likely to be a head principal as they are an assistant principal. No other group even comes close to this ratio. One possible explanation could be that this is just the way things are in Texas; that the ratios are reflective of the career ambitions of the individuals in the data. Alternatively, some might suggest that assignment to

administrative roles is based upon individual merit. Still others might conclude the disparity in ratios is a reflection of large scale social determinants which sort people into and out of certain roles within the education system.

Suppose the assumption is made that the only difference between the individuals represented in table 1.1 is their ethnicity. That is to say, they are all equal in measures of ambition, experience, leadership ability, and any other characteristic that could potentially set candidates apart in terms of fitness. Under this assumption, should one expect similar assistant-principal-to-principal ratios for all ethnicities? Would expecting otherwise equate to recognition of societal prejudice? Questions such as these reside at the core of this study.

The purpose of this study is to explore and interpret the disproportionate assignment of school leaders in Texas. Through a combination of descriptive and inferential techniques, I seek to explore inequalities across ethnicity and gender in the selection and career paths of Texas public school administrators beyond that represented by Table 1.1. I also explore how trends in leadership selection and distribution have changed over time. Finally,

I investigate what factors may explain the disproportionate representation of ethnicity and gender in administrative roles.

### **Research Questions**

The research questions driving the current research study are as follows:

1) *What is the current state of leadership production<sup>1</sup> and administrative job availability in the state of Texas and how have they trended over time?* Specifically, how many graduates are coming out of principalship programs each year? How many educators attain administrator certification? How many educators with principal certification procure an administrative position?

2) *How do the career paths of educators with principal certification differ across ethnicity and gender?* Specifically, are there differences across ethnicity and gender in the likelihood of being promoted to a higher level administrative position? Are there differences in the time it takes to procure initial administrative job

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<sup>1</sup> The term production is used to describe the amount of educational administration master's graduates coming out of Texas' public institutions of higher education.



placement after certification? In the time it takes to advance to head principal?

3) *What factors affect an administratively-certified educator's probability of being selected into a campus leadership position?*

### **Significance**

The significance of this study is two-fold. First, while the field of educational administration is gaining knowledge about administrators' careers in some states, we have little knowledge of Texas' unique context. Second, there has been limited exploration of the social processes<sup>2</sup>, either explicit or hidden, which drive the selection and promotion of school administrators. This study combines the analysis of Texas administrator career-paths with an exploration of the large-scale social determinants which may be at work in the selection and movement of campus leaders. This report not only can inform Texas policy, but may serve as a blueprint for other researchers to explore the production, distribution, and careers of public school administrators.

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<sup>2</sup> Phenomena such as homosociability – the process by which individuals in power replace themselves with candidates that most closely resemble themselves, typically in ethnicity, age, and gender. Homosociability and other social factors are explored in chapter two.

## **Preview of Literature to be Explored**

The literature reviewed in chapter two of this study come comes mainly from the disciplines of educational administration and organizational behavior. Many of the educational administration studies investigate the underrepresentation of females and minority educators in leadership positions (e.g. Joy, 1998). Despite encouraging gains over time in the representation of non-Whites and females in campus leadership positions, these studies reveal a persistent and ongoing overrepresentation of Whites and males in leadership positions (Gates, Ringel, Santibanez, Ross, & Chung, 2003; Lankford, O'Connell, & Wyckoff 2003; Gates et. al., 2004).

The second strand of educational administration literature reviews studies which seek to understand the social factors at play in the disproportional selection of campus leaders. These studies describe a propensity for the majority administrator group, White males, to replace themselves with more White males (Blackmore, Thomson, & Barty, 2006; Tallerico, 2000a). Homosociability and similarity-attraction theory are presented as major influences in the principal selection process. Other studies in this strand reflect the lived experiences of

minority and female administrators as they combat negative valuations of their competency to be campus leaders (e.g. Pollard, 1997).

The organizational behavior literature contains studies which look at the roles ethnicity and gender play in the evaluation and promotion of workers. These studies find that even when controlling for a variety of work-history factors, non-Whites and females are rated lower on annual reviews, and are promoted significantly less than their White and male counterparts (Greenhaus, Parasuraman, & Wormley, 1990; Landau, 1995).

Together, the aforementioned research strands serve as only a preview to the foundation upon which the remainder of this report builds. Many more studies are explored in chapter two. Several findings described in these studies are reflected in the analysis results outlined in chapters four and five of this report. Chapter five includes additional literature references which help to further explain and contextualize the findings of this study.

## **Method**

The analysis in this paper is purely quantitative in nature. The primary data source used in this study is a collection of panel and longitudinal data sets purchased

from the Texas Education Agency (TEA) containing information on the educational background, certification, demographics, and job assignments of all certified educators in Texas employed at any time between the 1991-1992 and 2010-2011 school years. These data were analyzed in combination with publicly-available data reported via Texas' Academic Excellence Indicator System (AEIS), which contains a variety of information about students, faculty, and accountability outcomes for individual campuses in the state.

Because much of the data analyzed in this report is considered to be representative of the full population of interest, as opposed to a sample, a discussion of the appropriate analytical approaches for population and sample data is provided. A rationale for the inclusion of this discussion is included in Chapter 3.

A variety of descriptive and inferential statistics are contained within this report. Survivor estimates are calculated to help illustrate differences in administrator career paths. Multinomial logistic regression is used to explore factors relating to the probability of attaining promotion into campus leadership positions.

## **Assumptions and Limitations**

The primary data source for this study is supposed to contain up-to-date, accurate information on all Texas public educators. The data-collection process employed to compile this data involves many people and is particularly prone to human error. In turn, this means the data is inherently incomplete and inaccurate.

Many of the assumptions and limitations confronting this report are wrapped-up within the quantitative methods I have chosen to employ. Each method has its own assumptions about the underlying data it is describing. Discussion of each method's assumptions is provided in chapter three.

As with any quantitative study, I am limited by my data. In this study I make assumptions about the relative qualifications and leadership abilities of separate administrators in comparison to one another based upon the data I have to describe them. That is to say, I make the assumption that the data available to me is sufficient to evaluate the potential leadership abilities of the research participants and that any remaining variability in leadership ability is randomly and evenly distributed across and within gender and ethnicity groups.

**Definitions**

An assortment of educational-administration-specific and author-created terms are utilized in this report to describe the roles of educators. The term campus leader is used to describe presently employed assistant principals and principals. Head principal is used to recognize only those who are the head administrator of record for a particular campus. This is not meant to include curriculum leaders or campus-within-a-campus leaders. That being said, the classifications of individual educators in Texas are only as accurate as their role is reported to the state. Interpretations of administrative roles could differ both within and between LEAs.

The term administratively-certified is used to describe educators with valid licensure to work as an assistant or head principal. The required certifications for such licensure have changed over time and the methods I used in this paper account for that. The term educator workforce is used to indicate all individuals with educator certification - this excludes any and all campus and district level staff whose jobs do not require certification by statute (e.g. office and support staff, maintenance, custodial, etc.)

The Texas Education Agency codes each educator's ethnicity into one of twelve categories: African American, American Indian/Alaskan Native, Asian, Black/African American, Hispanic, Hispanic/Latino, Native Hawaiian/Pacific Islander, Native American, Not Specified, Other, Two or more races, or White. For this paper, I have chosen to collapse these twelve categories into five: White, Black, Latino, Asian, and Other. There are a variety of reasons I have chosen to do this - some methodological, some due to convention. Seven of the original twelve categorizations account for less than 1% of all participants in the study. Calculating separate parameters for groups of such small size would not lend itself to robust findings. Of the other five groups, Black/African American, African American, Hispanic, and Hispanic/Latino, and White, I chose to do some combining. All Black/African American and African American educators were recoded into a single group (Black). I have chosen to use Black in my discussion and analysis, instead of African American because it can account for individuals who may identify as Black but not as African American (e.g. Africans, Caribbean Americans, etc.). All Hispanic and Hispanic/Latino participants were recoded into a single group (Latino).

Whites and Asians were left as already categorized in the data. The remaining categories were combined into Other.

I recognize there is no collection of ethnicity and gender categorizations that would satisfy the variety of manners in which all study participants would self-identify. I also recognize the contention that classifying individuals based on their ethnicity and gender only serves to further entrench these differences in our collective conscious. I rebuke this contention and believe the five ethnicity groups I have created for this study remain useful. It is my hope that gender and ethnicity categorizations as constructed in this study will serve as "entry points" for the creation of critical spaces in which to discuss our state's public education system (Calás, Smircich, Tienari, & Ellehave, 2010, p. 247). The ethnicity and gender indicators I rely upon for this study, while not exhaustive of the numerable intersectionalities lived by Texas educators, still serve an important role in understanding trends in the diversity of our public school personnel.

### **Organization of the Proposal**

Immediately following this introductory portion of the paper is chapter two: Review of Literature. It contains a



review and analysis of the research fields this paper seeks to advance. Specifically, I review relevant, large-scale studies of leadership distribution trends as well as related studies exploring public school administrator career paths. I follow this with a review of educational administration research about the selection of administrators and the social factors which drive their appointment and promotion. The chapter concludes with a review of organizational behavior literature.

Chapter three begins with a restatement of the motivating problem behind this study. I then provide a rationale for a quantitative approach in answering my research questions. The remainder of chapter three focuses on the data utilized in this study, the analytical methods I have chosen to employ, and the assumptions and limitations of said methods.

Chapter four contains the results of the quantitative analyses outlined in chapter three. When appropriate, summary tables are provided alongside the written description of each analysis' outcome. Implications are provided for certain outcomes of interests, along with tie-backs to the literature explored in chapter two.

Chapter five is a summary of the first four chapters of the study. It is in that chapter that I reset the study as a whole, provide analysis of the findings in chapter four, explore implications of the study's outcomes, and posit future directions for research.

### **Summary**

This paper provides a highly-detailed quantitative analysis of the production and selection of Texas public school administrators. This study explores the current state of leadership production in the state of Texas and how that production has trended over time. This study also describes how the career paths of Texas public school administrators differ across ethnicity and gender. Finally, I investigate what factors may be at play in the disproportionate selection campus leaders.

## **Chapter Two: Review of Literature**

This study seeks to evaluate the current and historical trends in administrator production and selection in the Texas public education system. Additionally, this study aims to understand what differences, if any, exist in the career paths of administrators of different genders and ethnic backgrounds. In order to provide some background and context to the analysis of this paper, I have chosen to divide this literature review into three sections: 1) studies that have explored underrepresentation and selection into campus administrative positions, 2) work which has sought to explain the social factors guiding these phenomena, and 3) studies from the discipline of Organizational Behavior.

### **Underrepresentation Studies**

Joy (1998) conducted one of the first, comprehensive, nation-wide explorations of underrepresentation in school administration. Her study used the Schools and Staffing Survey (SASS) from the 1987-88 school year and its 1988-89 follow-up report. Joy compiled descriptive statistics and created a variety of inferential models to determine the relative probability of teachers seeking administrative

positions in a single school year, to secure a campus leadership position in the next. Joy illustrated the underrepresentation of women in the public school principalship by explaining that while 86% of elementary teachers were female, they comprised only 22% of elementary administrators. At the secondary level, 51% of teachers were female yet held only 25% of the principal positions. Using the attainment of an educational administration master's degree as a proxy for one's desire to enter the principalship as well as a variety of other individual characteristics, Joy tested whether women's underrepresentation in school administration was discriminatory. She based her discriminatory judgment on the nature of control variables entered into her models - suggesting that all candidates were matched on relative job fit. Joy found that in all of the computed models, being male provided a distinct, significant advantage in probability to procure an administrative position. For the time covered in her study, the likelihood of male promotion ranged from 3.7 to 9.3 times greater than that of females.

A close review of Joy's work reveals limitations in her findings. First, attainment of an educational administration master's degree may not be the best proxy

for desire to enter campus leadership. In many cases, educators attain graduate degrees with no intention of leaving the classroom, but rather to secure an increase in pay. Joy used data covering a period of just two years from a nationwide sample in which just 90 individuals comprised the group of teachers promoted into campus administrative positions. Joy's work also was published nearly 10 years after the data she analyzed was originally collected.

A widely cited study using the same SASS data as Joy was authored by Riehl & Byrd (1997). I have elected not to include any interpretation of their findings because I deem their methodology unsound, but because of the article's prevalence<sup>3</sup>, it should not go unmentioned.

Beginning in 2003, a series of studies were funded in partnership between the RAND Corporation and the Wallace Foundation. The first of these to be published was authored by Gates, Ringel, Santibanez, Ross, and Chung (2003). The report found an upward trend in the percentage of females becoming administrators between the 1987-1988 and 1999-2000

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<sup>3</sup> The authors used logistic regression to determine the effects of various individual characteristics (experience, education, family structure, campus characteristics, etc.) upon a teacher's probability to enter an administrative position. Some of these individual characteristics were inappropriately conceived proxies built from variables in the original SASS. The authors also chose to exclude a combined model for men and women, thus quashing their opportunity to report and interpret any main or interaction effects a gender variable may have had. In all models, they found that probabilities for administrative job procurements amongst females were far lower than for males. According to Google Scholar, the article has been cited over 50 times since its publication.

school years. Despite this trend, females accounted for just 44% of public school administrative positions despite being 75% of the teaching force in 2000. The authors also found that during the same school year, members of a racial minority comprised only 11% of public school administrators despite the much larger and growing share of the national educator body being members of non-White ethnicity groups.

The second report of the RAND and Wallace collaboration was authored by Lankford, O'Connell, and Wyckoff (2003) and represented a synthesis of four reports on the public school administrator landscape in the state of New York. Using state-wide, comprehensive, longitudinal data similar to that of the present study, Lankford, O'Connell, and Wyckoff analyzed the career paths of some 700,000 teachers and administrators. Despite their assertion of "no evidence suggesting that females are less qualified than males to be teachers or school leaders", the authors found that once certified as an administrator, women were less likely to secure assistant principal positions (p. A-6). The report provides evidence not only of gender disproportionality in New York's administrative ranks, but perhaps even prejudicial selection of principals. Potential explanations for this gender

disparity come from an additional finding that 74% of administratively-certified men apply for campus leadership positions compared to just 64% of women. This difference in application rates could be a signal of lower administrative ambition amongst women. Men were also more likely to report being interviewed for an opening than are women, 13% and 7% respectively, however the relationship between this figure and job applications rates were not discussed in the report.

The third RAND and Wallace report was authored by Gates et. al. (2004) and performed a somewhat similar analysis on North Carolina data as Lankford, O'Connell, and Wyckoff (2003) did with New York State. Aside from analyzing school leadership in different states, the biggest difference in these reports is that Gates et. al. had a specific interest in diversity and equity within the administrative workforce. The authors found an increasing proportion of females securing administrative positions over the years leading up to their study's publication. Despite this encouraging finding the authors described the gender gap as "alive and well" explaining that "men were four times more likely than women to become principals ... and over three times more likely to become assistant

principals" (p. x-xv). Although the authors posited the cause for this discrepancy could stem from gender-based discrimination, they did not explore the subject at length. The authors also described an increase in the percentage of principals from an ethnic or racial minority, yet expressed concern the state may not be able to sustain the increase because of higher turnover rates for African-American educators than non-African-Americans in the leadership pipeline.

The final report to come from the RAND/Wallace collaboration was a profile of Illinois administrators authored by Ringel, Gates, Chung, Brown and Ghosh-Dastidar (2004) - a companion report to the Gates et. al. 2004 piece. The authors found that while the percentage of female and non-White administrators had increased over time, White educators remained overrepresented in campus leadership. Similar to North Carolina, the authors suggested that Illinois may struggle to maintain increasing number of minority principals; mainly because the population of Illinois teachers is slowly becoming less diverse and therefore narrowing the diversity of the administrator pipeline. Finally, unlike North Carolina, the



authors found race to have only a small relation with turnover rates for principals in Illinois.

### **Potential Causes**

Educational administration remains a male-dominated field, particularly in the secondary principalship and the superintendency (Gates et al., 2003; Tallerico, 2000a, 2000b; Young & McLeod, 2001). Given the previously described trend of the disproportional distribution of women and ethnic groups in administrative positions, it would be wise to explore potential causes. A review of the literature provides two main streams of explanation, 1) differences between male and female educators, and 2) social factors involved in the selection process. These streams are explored below.

#### **Differences between male and female educators.**

Joy (1998) offered three alternative explanations for the male-favoring, prejudicial promotion of teachers into administrative positions which she described in her analysis of SASS data<sup>4</sup>. These explanations are be used as a

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<sup>4</sup> These three explanations are the basis of this subsection because no other papers suggested inherent differences between male and female educators other than gender.

starting point for exploring potential causes of disproportionality.

Joy first hypothesized that female educators may view administration as a "man's game" and therefore curb their own ambition to enter leadership positions. In her quantitative analysis, Joy was unable to measure gender differences in ambition, but estimated "the likelihood that women with educational administration degrees are less likely than men with these degrees to desire promotions is small" (p. 201).

A qualitative study was performed by Young and McLeod (2001) which offers more perspective on the matter. The authors reviewed records of 127 female students affiliated with a single educational administration department of an Iowa university. From that group, 20 women were selected whom represented a variety of roles in the education system from elementary administration to the superintendency<sup>5</sup>. In their study, Young & McLeod defined career aspirations for female educators as "...a complex concept that involved

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<sup>5</sup> While this study does not consider superintendents and central office personnel, the individuals who occupy these positions within a district typically fall under the umbrella of educational administration and have similar educational and work backgrounds as do campus leaders. Therefore, this study and other which discuss the superintendency remain relevant to the literature review.

career commitments, positional goals, and leadership orientations" (p. 469). They determined that a woman's likelihood to pursue and enter administration was heavily dependent on three factors: type(s) of administrative role models, proximity to transformative leaders, and the support and encouragement they receive from their superiors. While promotion from a place of isolation was not deemed impossible, it was ideal levels of the three, aforementioned factors that steered most of Young & McLeod's participants into administration.

Joy's (1998) second non-discriminatory alternative explanation conjectured that women may not apply for campus leadership openings as often as men. Potential explanations for this discrepancy included responsibilities stemming from family responsibilities (i.e. having young children at home). Eckman (2004) also explored the role of family structure in the high school principalship and found that only 24% of female principals had children at home, concluding that "...either women with children at home are not applying for the high school principalship or they are not being hired" (p. 382). As mentioned in the discussion Lankford, O'Connell, and Wyckoff (2003) study on New York State, administratively-certified women's lower likelihood

to have reported applying for a campus leadership position may be a signal of lower ambition for an administrative career. These findings suggest that responsibilities such as child-rearing could be influencing both the rate at which women apply for administrative jobs, and the rate at which they are selected.

The final factor suggested by Joy (1998) is that men may simply make for better administrators. She suggests that males could be more productive administrators for a variety of social reasons, not the least of which being that students, teachers and the school board may be more respectful of their authority because they are men rather than a valuation of their training and qualifications (p. 201). This sentiment that men garner more respect and authority based simply on their gender plays into the entrenchment of ill-conceived social norms which favor men over women as school leaders. Further exploration of this and other social factors is offered in the following section.

#### **Social factors involved in the selection process.**

Pollard (1997) conducted a qualitative study with 20 Black urban elementary school principals as participants. 12 of the participants were female. The study asked these

school leaders to reflect on their past and present careers as school leaders and the challenges they have had to face, based upon their race. Pollard revealed that all of the study's participants felt they had to actively counteract notions of incompetency based simply upon their race. The participants felt their abilities and gifts were undervalued, underrated, and under-appreciated by parents, faculty, and central office. Should teachers and central office administrators question or even challenge the authority and competency of Black administrative candidates, it would not be unreasonable to expect lower levels of promotion and higher levels of turnover for Black school leaders. This would in turn lead to underrepresentation of Black educators in school leadership positions.

***Female educators from a racial minority.***

To this point in the literature review, the experiences of administrators from ethnic and minority backgrounds has not been discussed in detail. Nor have the experiences of administrators whom are both female and belong to an ethnic or racial minority. Consider the following quote:

Almost anyone who remembers "school days" has two images of school officials. The favorite teacher, in fact nearly every teacher, was probably a woman. But the feared and revered final authority, the principal, especially in high school, is likely to have been a man. When the memory then turns to the race of the principal, the pattern of the White, male school administrator begins to emerge. (Coursen, 1975, p. 1)

This very common recollection of one's own school days fuels the exclusion of race, gender, and other socially-constructed characteristics from the educational administration discourse.

In the same Pollard (1997) article mentioned earlier, one of the study participants who had been a principal for two years discussed how being an African American female affected her job. She described that some of her staff did not respect her authority because she was Black. On top of that, she said the following of her staff "[they] definitely don't want to take orders from anybody Black and female" (p. 361). This is particularly troubling in light of the gifts that other study participants felt they brought to the table as campus leaders. For example, two of

the respondents felt that their leadership of a majority White faculty allowed them to bring a deeper knowledge about interacting with the majority African American student population. When non-White leaders are filtered out of schools with non-White student populations, districts risk keeping advantageous perspectives and skill sets from benefitting the children who need them the most.

Fitzgerald (2003) explains the error in lumping together the experiences of women from different ethnic backgrounds and assuming they have a singular, shared experience. She states that "women from ethnic groups other than White are a minority in a minority setting" (p. 433). Because of this, feminist educational administration research often fails to distinguish *women* from *White women* making Whiteness a taken for granted norm (p. 433). In other words, the varying intersections of race with the female gender are often ignored in the research of our field. This results in the experiences of White female educators who have their own race-based privilege being conflated with the experiences of individuals at other intersections of ethnicity and femaleness. Fitzgerald suggests that:

Knowledge production for and about educational leadership needs to be dismantled to provide a standpoint from which to theorize and research the realities of leadership through the experiences of women from a variety of ethnicities that simultaneously encourages and permits the situating of such knowledge and action in the cultural spaces in which they arose ... this includes acknowledging and interrogating Whiteness as a specific privilege and taken-for-granted construct (pp. 441-442).

Dillard (2010) shares Fitzgerald's sentiment in her exploration of the intersectionality of race and gender experienced by African American females in the field of education. Dillard makes the case that traditional powers imbalances have curbed the legitimacy and prevented the emergence of research on the experiences and identify of African American, female educators. Dillard calls for the establishment of an "endarkened epistemology" as a means to shed the exclusory forces that surround research on the experiences and contributions of women of color in education. She uses the term endarkened (an allusion to skin color) to refer to the explicit recognition and consideration of race in the interpretation and construction of research. An endarkened epistemological approach can inform our collective understanding of the careers of minority female administrators.



Further still, there is error in assuming female leaders have a shared, singular experience even within a particular ethnicity. Reed and Evans (2008) performed a detailed qualitative analysis of the experiences, leadership practice, and beliefs of African-American, female administrators in a predominately-Black, urban school. Their study began with a review of two major assumptions common in educational administration literature: one, that African American principals "tend to identify/empathize with African American students", and two, that African American female principals are nurturing leaders, caregivers or 'othermothers' (p. 490). In their case study, Reed and Evans found that neither of these assumptions held for one particular participant. The principal not only had an antagonistic relationship with her students, but carried a deficit view of the African American community surrounding the school.

***Homosociability and gatekeeping.***

Blackmore, Thomson, and Barty (2006) apply the concept of homosociability (1984) to the selection of school leaders. They describe homosociability as the tendency of decision-makers in the administrator selection process to pick candidates like themselves. Given the proclivity for

higher proportions of males to ascend to positions influential in the administrator selection process (Gates et al., 2003; Tallerico, 2000a, 2000b; Young & McLeod, 2001), homosociability has the tendency to create a 'closed circuit' in which men replace men as school leaders (Blackmore, Thomson, & Barty, 2006, p. 312). These closed circuits work to exclude females and educators of color.

Homosociability was likely a factor driving the construction of a purported state-wide administrator shortage and subsequent policy creation intently critiqued by Young (2003). In her review of publications from state task forces (primarily comprised of White men) on the shortage of school administrators combined with the data from a previous study, Young was able to clearly illustrate the gendered construction of the 'shortage crisis' confronting the state. Even when confronted directly with the notion, little credence was afforded from the task forces to the role and experiences facing females in the state's administrative pipeline, much less their collective potential as a solution to the 'crisis'.

Tallerico (2000a) conducted a thorough examination of the superintendent search and selection process providing many findings relevant to the current research project.

Tallerico interviewed 75 individuals including educators seeking superintendency positions, members of superintendent selection committees, headhunters, and school board members from across the state of New York – all with the intention of understanding what individuals from these different groups “do to both help and hinder diversity in candidate pools and the superintendency” (p. 29). Tallerico employed gatekeeping theory which states that individuals seeking promotion into leadership positions must pass through a series of ‘gates’ which can be thought of as decision points in a process where candidates are either rejected or move toward the next gate. Gates are not confined to the individual level; they exist at the institutional level where rules and norms determine the progress of potential candidates. While she doesn’t use the term, homosociability plays a powerful role in gatekeeping. Quality candidates were deemed as those with prior experience at secondary campuses and in district leadership. Tallerico (2000a) explains that such a construction of quality disadvantages women and people of color, as such positions are traditionally held by White males.

Participants in Tallerico's (2000a) study agreed that the face-to-face interview portion of the selection process was the most influential factor in the hiring process. Candidates, headhunters, and board members alike cited the importance of positive chemistry and a good "gut feeling" coming from interviews as very influential in final decision-making. Tallerico suggests that the "hypervaluing of 'how we connected with the candidate' and 'who we could relate to best' is more likely to disadvantage people of color and females than White male applicants" (p. 36). One reason she cites is because so many of the gatekeepers in the selection process are White men. The other reason is similarity-attraction theory, a concept very similar to homosociability, in which feelings of interpersonal-connection between candidates and interviewers can create subconscious preferences amongst gatekeepers to bring aboard individuals like themselves. When the gatekeepers are themselves primarily White men and employ narrow definitions of quality candidates, closed-circuits emerge and a protected network of similar leaders continues forward.

In summary, the experiences, beliefs, and attributes, of female educators as well as the expectations placed upon

them, differ not just from men, but also differ between and amongst the various ethnicity groups these women come from. Despite these differences, gatekeepers rely heavily on preconceived notions and stereotypical expectations of the ability for administrative candidates to lead based upon what they see on the outside. These notions and expectations fuel the disproportionate selection and promotion of females and minorities in school administration, ignoring the unique perspectives and abilities that individuals from diverse backgrounds bring to the table.

### **Organizational Studies**

Greenhaus, Parasuraman, and Wormley (1990) performed a study on three large organizations from the banking, communications, and electronics industries. The authors constructed two matched groups of Black and White managers within these organizations totaling 1,628 individuals in managerial positions. These individuals and their supervisors were administered surveys to collect a variety of responses about the workplace. Examples of responses included job satisfaction and acceptance in the workplace for the managers, while their supervisors were asked to assess the performance and promotability of the managers.

The authors found that Black managers, even when controlling for age, organizational tenure, job function, and level within the organization were systematically more likely to hit a career plateau - seven years or longer in the same position. The authors also found that Black managers were systematically rated lower by their supervisors than were their matched, White counterparts. The authors were hesitant to claim bias as the main cause behind their findings, but suggested it as a possible reason for the discrepancies between Blacks and Whites in the survey results. This study did not look at the relationship gender played in the survey responses.

Landau (1995) conducted a similar study, but with 1,268 participants from a single Fortune 500 company. Landau analyzed survey responses from the participants and reviewed personnel records from the organization. The personnel records contained yearly performance appraisals which rated the study participants between low, average, and high potential for promotion in the company. Controlling for age, education, organizational tenure, salary grade, role, and even participant-rated satisfaction with the support they receive in their career, regression models found that females and non-Whites consistently and

significantly rated lower for promotional potential. Unlike the Greenhaus piece, this study did look at interactions of race and gender, but found no significant relationships.

Similar findings emerged from a 2012 meta-analysis of promotion potential studies, conducted by Roth, Purvis, & Bobko. The authors found that women, even when rated *higher* in job performance than their male counterparts, were given lower promotability ratings. When considered together, these studies illustrate that even when the relative merits of candidates for promotion within an organization are considered, difference between individuals defined by their ethnicity and gender still play a role in their upward mobility. These studies suggest that gender and ethnicity-based biases play a significant, damaging role in the selection and career paths of non-White and female workers.

### **Summary of Sources**

The literature reviewed in this chapter covers three main areas; 1) studies highlighting the disproportional selection of Whites and males into administrative positions at both the state and national level, 2) studies exploring the social factors at play in the underrepresentation of minorities and women in campus leadership positions, and 3) discussion from the field of Organizational Behavior. These

sources have shown positive trends highlighted by the increased presence of minorities and women in leadership roles, but also illustrated the ever-present dominance of Whites and males as school leaders.

A variety of explanations for prejudicial selection of administrators was offered. These explanations included a gatekeeping process heavily influenced by homosociability and selectivity-attraction theory. They also included evidence for gender and ethnicity-based bias in the evaluation of individuals within organizations. These theories should not be considered in isolation, nor should they be considered as evidence of simple prejudice. They should instead be considered collectively as examples of the societal ills which combine to form the exclusionary forces of sexism and racism which pervade our education system. These forces are exemplified in the systematic processes by which non-Whites and females continue to be filtered out of school leadership positions.



### **Chapter Three: Methodology**

This research project has three main goals: 1) to describe the past and present production of school leaders in Texas, 2) to understand how the careers of administratively-certified educators differ for individuals of differing ethnicity and gender, and 3) to investigate factors contributing to the disproportional assignment of school leaders based on ethnicity and gender. This chapter begins with a rationale for using quantitative methods to answer my research questions. The remaining sections are dedicated to describing the data used in this study, the methods I have chosen to analyze the data, and a discussion of the assumptions and limitations of those methods.

#### **Need for Quantitative Study**

The research questions of this study, particularly those interested in how leadership preparation and placement has trended over time, lend themselves to quantitative exploration. Descriptive statistics, summary tables, and a handful of statistical models provide a detailed explication of Texas' unique leadership landscape and the changes it has undergone over the past twenty years. The last research question which centers on factors

affecting selection practices, could certainly benefit from supplemental qualitative data, however the richness of the data set utilized for the present study allows me to compile enough unique characteristics for each administratively-certified educator that I can make conclusions about factors contributing to the probability of being promoted into campus leadership.

### **Sources of Data**

This study uses a multitude of databases compiled by the Texas Education Agency (TEA). Some of the data was purchased from TEA and is managed by the University of Texas at Austin's Principalship Program (UTAPP) for whom, at the time of the study, I am a research assistant. Other data come from Texas' publicly available Academic Excellence Indicator System (AEIS), which annually releases comprehensive staff, student, and accountability data aggregated at the campus level.

The purchased data contains information compiled by the State Board for Educator Certification (SBEC), a division of TEA which manages the certification information on all public school educators. The certification information lists the effective and expiration dates for each certification gained over an educator's career, as

well as the specific subject area and grade levels in which they are certified. The purchased data also contains specific information on the job assignment(s) of all employed educators for each school year between 1991-1992 and 2010-2011. The job assignment information lists each individual's role<sup>6</sup>(s) as well as the name of the campus, district, and region<sup>7</sup> in which they work. The final portion of the purchased data contains demographic information on each of the participants in the study. This data tells indicates individuals' birth date, ethnicity, and gender.

### **Site**

All data analysis took place within the state of Texas - primarily on campus at the University of Texas at Austin.

### **Participants**

The participants in this study include all certified educators employed in the public school system at any point between the 1991-1992 and 2010-2011 school years (N=985,232). Of these individuals, nearly all analysis is performed upon those, who at any time were administratively-certified (93,424). To provide an idea of

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<sup>6</sup> e.g. classroom teacher, assistant principal, principal, etc.

<sup>7</sup> All school districts belong to one of 20 regional education service centers (ESC) constructed by TEA. Figure A.1 of the appendix contains a map displaying the boundaries of each ESC.

how these 93,424 educators were dispersed over the twenty years of the study, the total number of them employed as campus administrators in any given school year ranges from a minimum of 10,165 to a maximum of 17,591.

Individuals included in this study were passive participants. No educators were contacted for any additional data collection. All the data used for this study comes from the existing data sets purchased from TEA and obtained through AEIS.

### **Analyzing the Data**

The data from SBEC is in person-period format. This means there are multiple "rows" for individual educators. These separate rows contain data about a unique certification occasion. These occasions can mark an educator's initial entry into the SBEC system, they may be reflective of an educator adding a new certification area to their license, or could even be a renewal of an existing certificate.

The TEA data containing responsibility and assignment information for all educators is split across 20 different files. Each file marks the campus and role information each individual educator was assigned to during a particular school year. Because the files represent a single, uniform

period in time for all individuals described within, they are considered panel data. Many educators have multiple rows in the data set to account for differing grade level and subject assignments. Some educators even have multiple campus assignments.

One of the major tasks in the data analysis portion of this study has been the clean-up, sorting, and recombination of the disparate licensure, assignment, and demographic files. Thankfully, the state uses unique identifiers for all educators and campuses in the system. This allowed me to identify schools and individuals across multiple data sources. Below is a listing of the variables used in this study, how they are measured, and how they were used (i.e. as dependent compared to independent variables).

**Table 3.1 - Study Variables**

Variable	Nature of Measure	Use
Ethnicity	Categorical	Independent
Gender	Categorical	Independent
Age	Continuous	Independent
Region	Categorical	Independent
Role	Categorical	Dependent
Time <sup>8</sup>	Continuous	Independent
Experience in education	Continuous	Independent
Tenure with district	Continuous	Independent
School characteristics <sup>9</sup>	Primarily continuous	Independent

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<sup>8</sup> A more thorough description of the use of time in this study is offered in table 3.2

## **Descriptive Statistics**

The population of interest for this study is all administratively-certified educators who have worked in Texas public schools over the past twenty years. The data used for this study covers that entire population. Because I have population level data at my disposal, there is no reason for me to exclusively generate inferential statistics. Descriptive statistics and simple mathematical procedures allow for the calculation of the actual population parameters in many cases. Estimating real population parameters with population data (N as opposed to n) precludes reliance upon the determination of significance (p values) in statistical findings that is so prevalent in modern research.

Below is a listing of the measures to be explored, the descriptive methods to be employed, the variables involved, and the research question they address:

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<sup>9</sup> e.g. percentage of students economically disadvantaged, type of school (elementary, middle, high, other), AEIS accountability rating, etc. Most of these variables are percentages describing students characteristics and are therefore continuous. Others such as school type and accountability are categorical.

**Table 3.2 – Descriptive Methodology**

<b>Outcome of Interest</b>	<b>Output / Method</b>	<b>Variables considered</b>	<b>Research question addressed</b>
Descriptions of all subjects across ethnicity and gender	Tabulations, percentages	All	N/A
Production of educational administration master's degrees	Raw counts, percentages, ratios, graphs	ethnicity, gender	1
Production of first time administrator certificates [*19 – one for each year between 1992 and 2010] <sup>10</sup>	Raw counts, percentages, ratios, graphs	ethnicity, gender	1
Assignment of first time administrators (1st AP or prin job) [*20 – one for each year in data]	Raw counts, percentages, ratios, graphs	ethnicity, gender	1, 2
Educator role by ethnicity (teacher, ap, or prin) [*20 – one for each year in data]	Tabulations (raw count & percentage), graphs	ethnicity, gender	1, 2
Educator role within ethnicity (teacher, ap, or prin) (teacher, ap, or prin) [*20 – one for each year in data]	Tabulations (raw count & percentage), graphs	ethnicity, gender	1, 2
Time 1 – Tenure in education system	Mean	ethnicity, gender	1, 2
Time 2 – time between system entry and administrative certification	Mean	ethnicity, gender	1, 2
Time 3 – time between administrative certification and initial administrative placement	Mean	ethnicity, gender	1, 2
Time 4 – time between first assistant principalship to head principalship	Mean	ethnicity, gender	1, 2
Time 5 – time between administrative certification and head principalship	Mean	ethnicity, gender	1, 2

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<sup>10</sup> For all outcomes measured separately for each of the twenty years of data, a variety of visual aids are constructed to show how trends have changed over time.

Time 6 - time between system entry and any administrative placement	Mean	ethnicity, gender	1, 2
Time 7 - time between system entry and head principal placement	Mean	ethnicity, gender	1, 2

### **Survival Estimates and Probabilities**

In only a few instances do I employ quantitative methodology that departs from frequencies, tabulations, means, and other related descriptive methods. These instances are the calculation of a) relative hazard ratios between groups of individuals, and b) relative probabilities between individuals in the data set.

Event history analysis is a collection of methods used to understand whether and when events occur over a given period of time, with particular interest paid to changes in status (e.g. an individual being hired into a campus leadership position). In this study, event history analysis provides another layer of understanding to the differences in career paths experience by members of differing gender and ethnicity groups.

The specific method to be employed is the estimation of Kaplan-Meier survivor functions. Conducting these analyses requires the construction of time sensitive data sets containing measures of time passage between certain



major events in educators' careers. The outputs of these models allow us to understand changes in the average risk faced by the various intersectionalities of ethnicity and gender for experiencing the event of interest - in this case, procuring a campus leadership position. In other words, we can understand the differing rates at which administratively-certified educators exit the classroom and enter campus administration, as well as how those rates change over time and how they differ across ethnicity, gender, and their various intersections. A detailed explanation of survival estimates is offered in chapter four.

Relative probabilities were constructed to determine how an individual's characteristics (specifically: ethnicity, gender, ethnicity\*gender, a variety of characteristics describing the campus in which they are employed, age, experience, tenure, role, region of state) affects their probability to remain a teacher, become an assistant principal, or become a head principal. These three outcomes are nominal in nature; therefore a multinomial logistic regression is ideal way to calculate these relative probabilities.

The characteristics listed above as covariates for the multinomial logistic regression models were chosen to provide rich detail about the work history and work environment from which the administrative candidates in this study come from. The intent is to control for the effects that a leadership candidate's individual characteristics, work history, and work environment might have upon their probability of being selected into an administrative position. Interpreting the effects of these variables in the model outputs gives us an estimation of how much a particular covariate (e.g. ethnicity) has upon one's probability of being hired as a campus leader, when holding constant the effect of a variety of other covariates (e.g. tenure with current district, previous year campus accountability rating, percentage of students on the campus that are economically disadvantaged, region of the state, etc.).

Multinomial logistic regression fixes the independent variable coefficients of the baseline outcome, in this case being a teacher, to 0. Model results allow the researcher to view the independent variable coefficients of the other outcomes (becoming an AP or becoming a principal) as changes in odds in relation to the base outcome.

Multinomial logistic regression assumes the independence of irrelevant alternatives (IIA). This assumption states that a model is not excluding additional alternative outcomes that would significantly change its output. McFadden and Train (1977) explain that incorrectly assuming IIA in a multinomial logistic regression will lead to inaccurate probabilities being reported. In the same article, the authors offer strategies for testing the presence of IIA. This study employs the commonly utilized Hausman and Small-Hsiao tests of the IIA assumption. Finally, consideration of variable homoscedasticity is not needed in multinomial logistic regression. This is because the identification technique of the model fixes residual variance to a generalized extreme value distribution.

As mentioned previously, the multinomial logistic regression models measure the effect of a variety of independent variables on an administratively-certified individual's role assignment. In order to determine whether or not certain independent variables "matter", p values are interpreted as a measure of statistical significance. Even though this study utilizes population level data, interpretation of these values is a necessity to determine the relative effects of said independent variables.

Covariates with p values of less than 0.05 are considered statistically significant.

Below is a listing of the probability and hazard models used in the study, as well as the variables included and the research question they address.

**Table 3.3 – Inferential Methodology**

Outcome of Interest	Method(s)	Variables considered	Research question addressed
Survival rate of teachers with “failure” defined as procurement of assistant principalship	Kaplan-Meier survivor functions	ethnicity, gender, ethnicity*gender, multiple characteristics of the campus in which they are employed, age, experience, tenure, role, region of state	2
Survival rate of teachers with “failure” defined as procurement of head principalship	Kaplan-Meier survivor functions	ethnicity, gender, ethnicity*gender, multiple characteristics of the campus in which they are employed, age, experience, tenure, role, region of state	2
Relative probability of being a teacher, ap, or principal [*20 – one for each year in data]	Multinomial logistic regression	ethnicity, gender, ethnicity*gender, multiple characteristics of the campus in which they are employed, age, experience, tenure, role, region of state	3

## Chapter Summary

In this chapter I have made the case for a primary reliance on descriptive statistics. The data being used for this study is stored in a variety of formats, but all together, they cover the entire population of interest.

Therefore, descriptive statistics create population parameters, rather than estimated statistics. I have also introduced Kaplan-Meier survivor estimates which help illustrate differences in careers paths for individuals at the various intersections of ethnicity and gender. Finally, multinomial logistic regression was described as the method for investigating which factors contribute to an individual's probability to be selected into an administrative position.

## **Chapter Four: Analysis & Findings**

This chapter is separated into two parts. The first is a report of all descriptive statistics outlined in Table 3.2. These descriptives form the responses to research question one, which sought to explore trends in administrator production and selection. The latter half is a report of inferential statistical models, all performed via a frequentist approach. The inferential models form the responses to research questions two and three, which dealt with differences in career paths based on ethnicity and gender, as well as the factors related to underrepresentation. For some models, particularly the multinomial logistic regression models, only representative findings are presented because the full model outputs are far from succinct. Full reports from all models can be found in table A.8 of the Appendix. All data management and statistical analysis for this chapter was performed using Stata statistical analysis software<sup>11</sup>.

### **Descriptive Statistics**

As mentioned in Chapter Three, this study utilized 985,232 participants representing any certified Texas

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<sup>11</sup> Stata IC ver. 11.2 64 bit

public school educator employed between the 1991-1992 and 2010-2011 school years. The total number of public school educators grew by roughly 12,000 each year from a low of 279,797 in 1991-92 to a high of 461,375 in 2009-10.

Gender representation in the public school workforce was virtually static over the 20 school years covered in the study at an 80%-20% split, females to males. The ethnicity of Texas public school administrators however, changed greatly over the same time frame. With a high of 73.9% of the workforce in the 1991-92 school year, White representation consistently fell each year to a low of 62.1% in 2010-11. Asians have accounted for a low of 0.61% of public school educators (n=1,705) in 1991-92 to a high of 1.19% (n=5,505) in 2010-11. The share of Black public school educators in Texas has risen gradually over the past 20 school years from a low of 8.47% (n=24,708) during the 1992-93 school year to a high of 10.36% (n=47,770) in 2011-12. Latino educators have seen a very substantial and steady increase in their representation over this same time period - from a low of 16.11% (n=45,087) in 1991-92 to a high of 25.05% (n=115,575) in 2010-11. The number of new, White educators coming into the workforce each year appears to be leveling off, where all other ethnicity groups are

growing each year. This suggests that White's share of the workforce will decline quite precipitously; especially should the number Latino educators entering the workforce each year stay on its current trajectory. Table A.1 of the appendix contains complete cross-tabulations of the participants' ethnicity and gender for each of the 20 school years covered in the study.

### **Research question 1.**

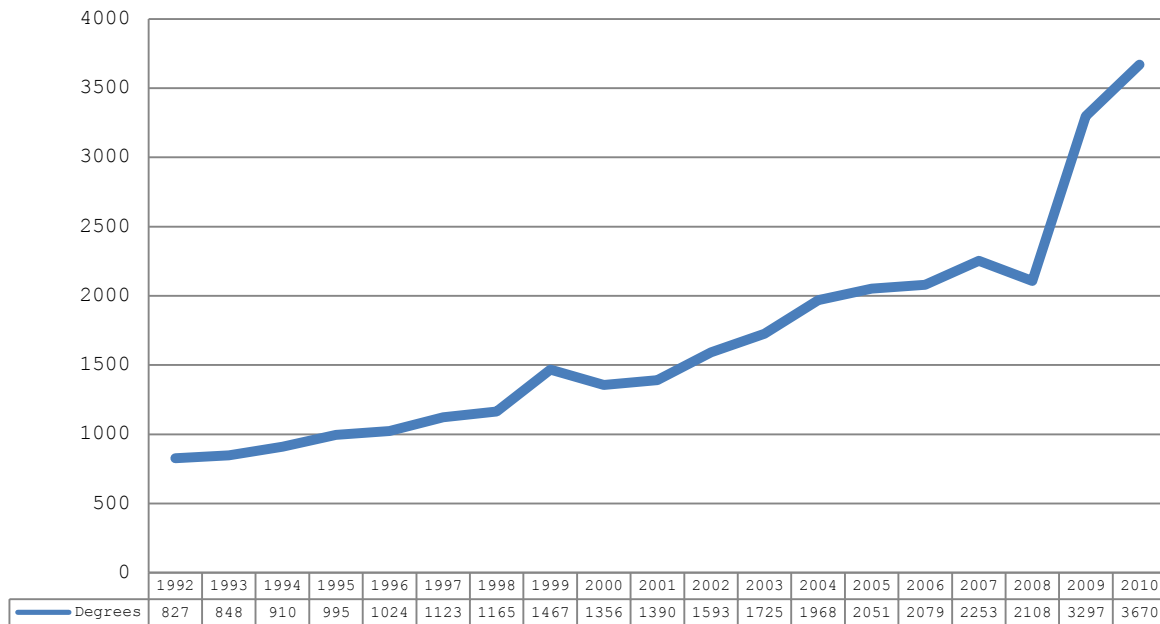
Shifting the focus from a broad picture of all educators toward campus leaders, recall that the first research questions asked "What is the current state of leadership production in the state of Texas and how has production trended over time?". To answer this question I proposed to quantify, for each of the 20 years in the study, the number of educational administration master's degrees awarded, the number of first time administrator certifications granted, and the number of educators entering campus leadership positions for the first time in their career.

Beginning with degree production, Figure 4.1 shows the number of educational administration master's a degrees awarded each year by public institutions of higher education in the state of Texas. Master's degree production



has seen a steady increase marked by a very sharp jump in recent years.

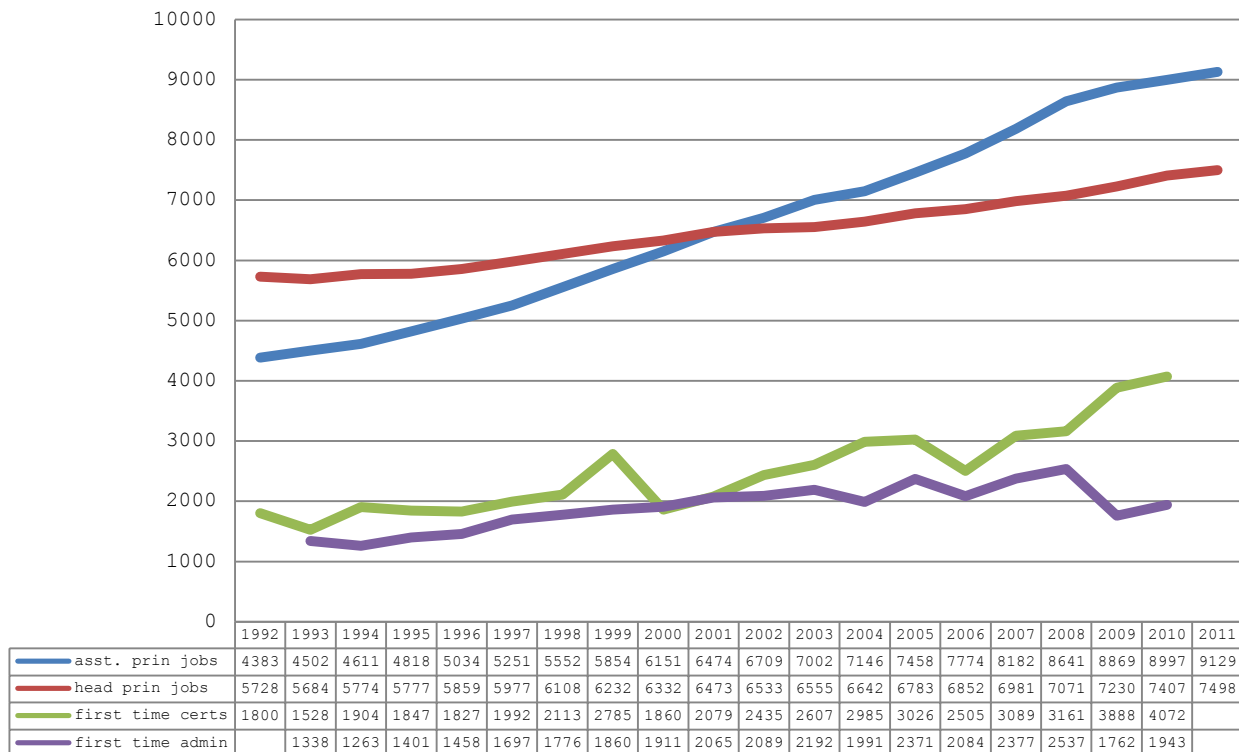
**Figure 4.1 - Educational Administration Degree Production by Public Institutions in Texas**



To provide some context to the production of first time certifications and procurement of initial administrative job placement, Figure 4.2 shows these figures along with the total number of campus leaders in the assistant principal and head principal positions across the state. As can be seen below, the number of first time administrative certifications fluctuates annually, but shows an upward trend over time. The same is true for the number of educators procuring first time administrative job placement. Another interesting trend from Figure 4.2 is

seen in 2001. During this year, the total number of assistant and head principalship positions in the state became equal. Since that time, there has been continuous, positive growth in the number of assistant principal positions, while growth in the head principalship has been modest (increases in total head principal positions can only come from new school constructions). This suggests that LEAs are electing to employ more assistant principals on their campuses.

**Figure 4.2 – Total Counts of Campus Leadership Positions and First Time Administrative Job Placements**



The data in figure 4.2 become especially interesting when the sharp increase in degree production beginning in 2009 (illustrated in Figure 4.1) is contrasted against the very modest growth of first time administrative job procurement over the same time period. This suggests the possibility of a flooded administrator market which will be discussed in chapter five.

Together, Figures 4.1 and 4.2 show that administrator production has increased steadily over the past twenty years with a sharp increase in the awarding of educational administration master's degrees occurring in the 2008-09 school year. The number of educators appointed to campus leadership positions for the first time in their careers has steadily increased over time at a rate very similar to that of first time administration certifications. In 2009, these trends change significantly as the number of first time administrator certifications jumped up 23%, while the number of first time campus leader hires fell 30%. Over the same twenty years, growth in public school leadership positions grew steadily, but not at a rate comparable to the growth in educational administration master's degrees - particularly in the years since 2009. In sum, trends in the growth of educational administration graduates, campus

leadership certifications, and first time campus leadership assignments have been steadily positive, with the exception of 2009 onward. In chapter five, explanations for these changes will be discussed.

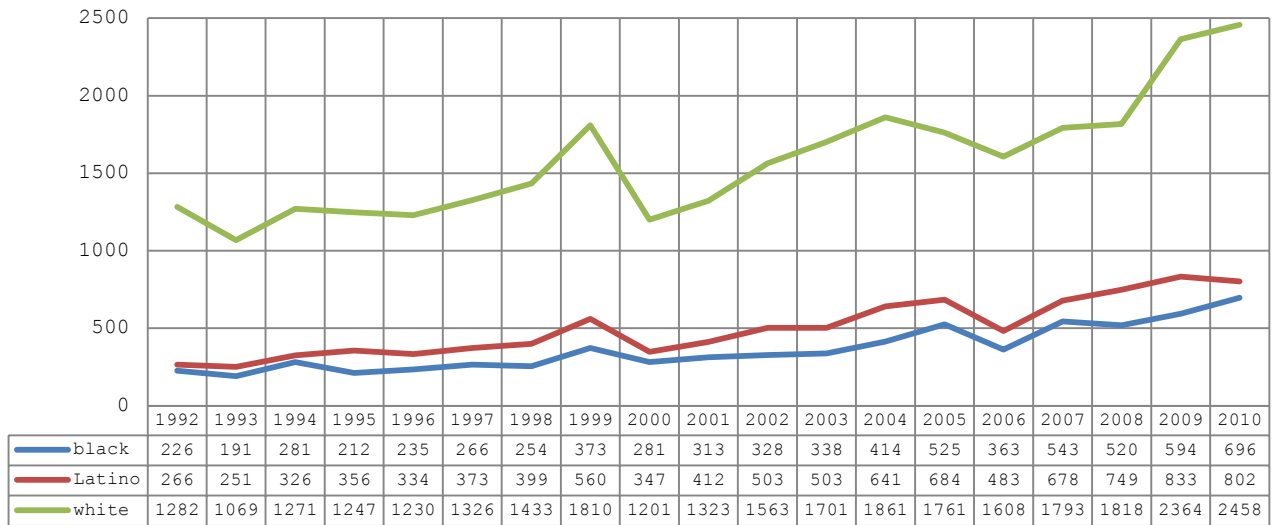
### **Research question 2.**

The second research question asks how the career paths of educators with principal certification differ across ethnicity and gender. I proposed a variety of methods for answering this question, both descriptive and inferential. The first step to answering this research question requires a thorough review of the production and certification of leaders over time and across ethnicity and gender. Subsequent steps included examination of the distribution of individuals across and within campus roles, as well as a comparison of measures of time between significant career changes - all disaggregated by gender and ethnicity.

### ***Production and certification.***

Figures 4.3 and 4.4 display the number of first time administrator certifications issued by the state, disaggregated by ethnicity and gender, respectively. These figures exclude certification renewals. Figures for Asians and others have been excluded from the chart as they are indistinguishable from the chart's baseline of zero.

**Figure 4.3 - New Certs by Ethnicity**



**Figure 4.4 - New Certs by Gender**

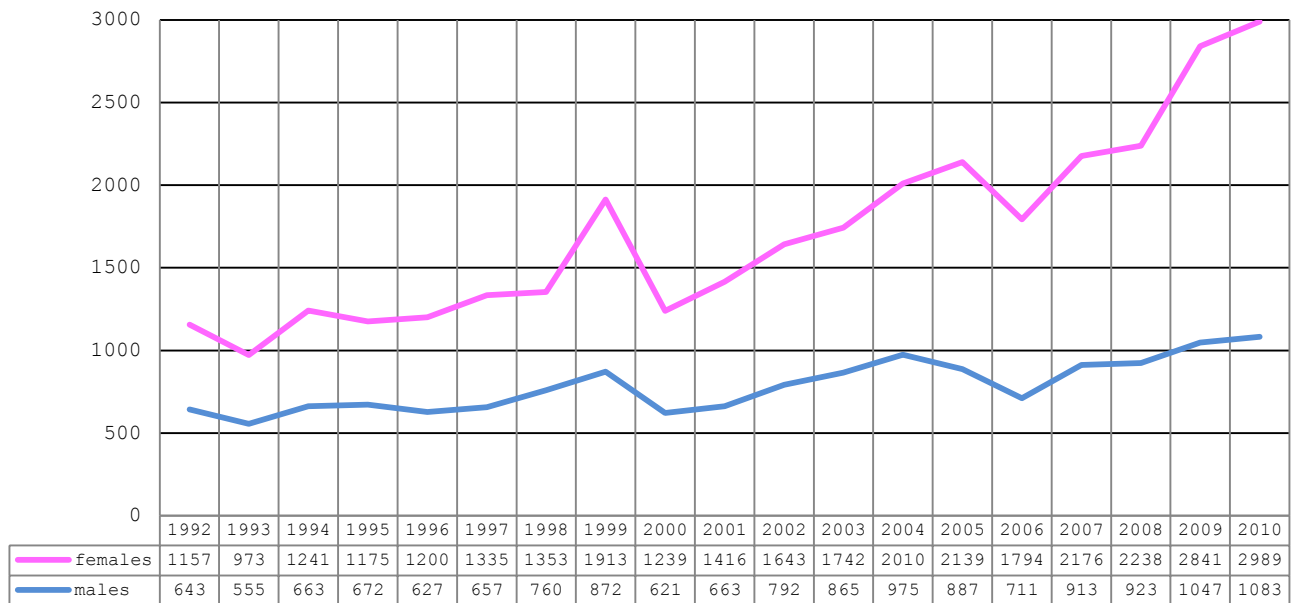
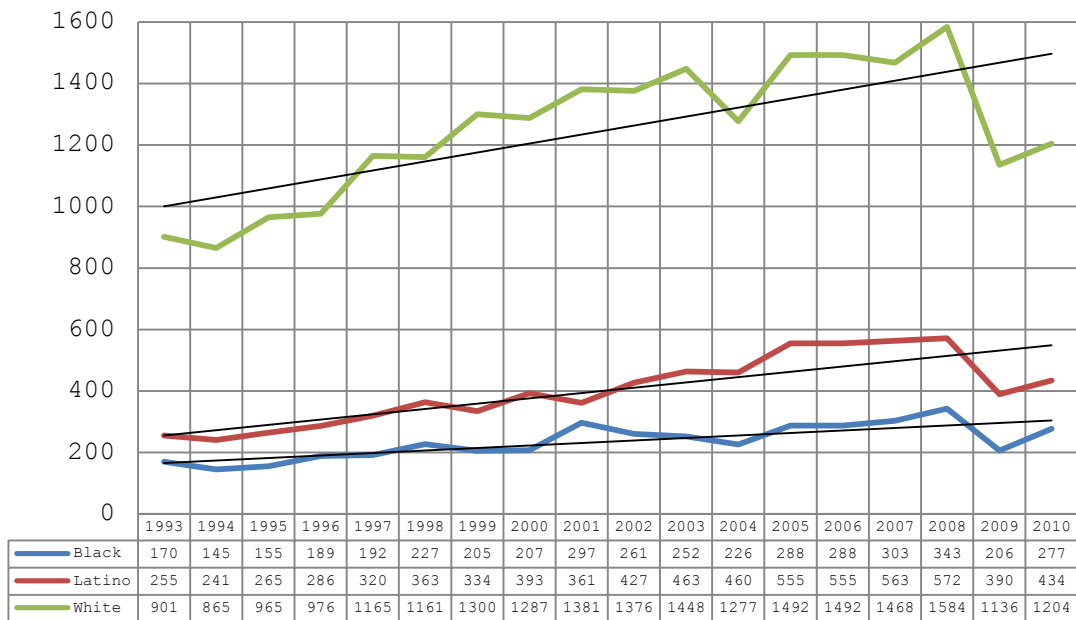


Figure 4.3 shows a steady and comparable rise in first time certifications across gender and over time with the

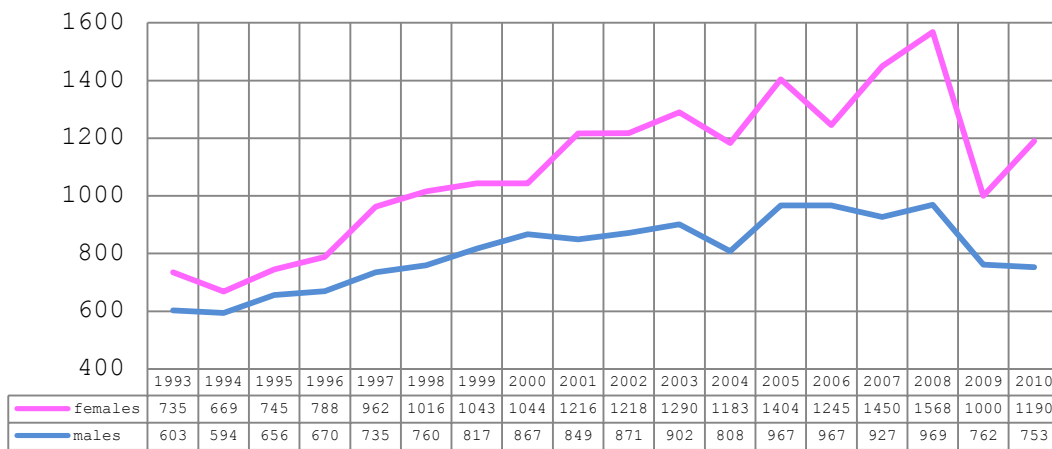
exception of 2009 and onward. Here we see a steep raise the number of first time principal certifications being procured by White educators. Figure 4.4 displays a widening gap over time in which women earn an increasingly larger share of first time administrative certifications. Similar to figure 4.4, there is a large upward bump in certifications beginning in 2009. This suggests that an unusually large amount of White women are securing administrative certification beginning in 2009 and moving onward.

Figures 4.5 and 4.6 display the number of occurrences in each school year that an administratively-certified educator procured their first-ever campus leadership position, disaggregated by ethnicity and gender, respectively. Figure 4.5 suggests nearly parallel growth in first time administrative job procurement between Whites and Latinos, while growth amongst Black educators was slower. Just as in Figure 4.2, there is a steep drop off in 2009 for the number of educators getting their first administrative job.

**Figure 4.5 - New Admin Jobs by Ethnicity**



**Figure 4.6 - New Admin Jobs by Gender**



After beginning at virtually the same level in 1993, figure 4.6 shows an increasingly widened gap over time for the procurement of first time administrative jobs between men and women. The year 2009 marks a very large drop off (-

568 or -36.2% drop) in the number of new administrative assignments for women. A similar, although far less severe drop occurred for male educators (-207 or -21.3% drop). A potential relationship between these 2009 onward changes in job securement and certificate production is explored in chapter five.

***Distribution across and within campus roles.***

In order to determine ethnicity and gender representation both within and across campus roles, I performed two separate cross tabulations for each of the 20 school years in the study - one with gender and role, the other with ethnicity and role. Individual school year cross tabulations were performed only upon educators with leadership administration certification valid during that particular year and working in one of three roles: 1) classroom teacher, 2) assistant principal, or 3) head principal.

Since the 1991-92 school year, White representation in campus leadership and teaching positions amongst the administratively-certified has dropped consistently over the years. At the highest, Whites accounted for 68% of assistant principals, 76% of head principals, and 73% of teachers - those numbers dropped to 56%, 66%, and 63%



respectively for the 2010-11 school year. Non-White representation has grown significantly and consistently over that same time frame. Administratively-certified, Black educators accounted for an all-time high<sup>12</sup> of 16.51% of assistant principals and 10.92% of head principals during the 2009-10 school year - up from 12.44% and 8.34% respectively in 1991-92. Administratively-certified, Latino educators account for 25.57 and 21.43 percent of assistant principals and head principals, respectively - up from a low of 18.36% and 15.21% during the 1991-92 school year.

Changes in gender representation tell a similar story. In 1991-92, females held 53.68% of administrative certifications while males had 46.32%. Female representation amongst this group has consistently risen since that time. In the 2010-11 school year, females accounted for 65.58% of certified educators in campus leadership positions in the classroom, while males accounted for the remaining 34.42%. Despite females holding the larger share of certifications, in 1991-92, they accounted for only 48.74% and 41.33% of assistant and head principal positions, respectively. Those figures have

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<sup>12</sup> This includes 2011-12 data

steadily risen over time to an all-time high in 2009-10 with females accounting for over 63% of both assistant and head principal positions.

In 1991-92, males with administrative certification were far less likely than women to remain in the classroom. In fact, 35.1% of administratively-certified males were in the head principal position at that time, compared to just 21.34% of females. In that same year, 58.82% of administratively-certified females remained in the classroom while 59.28% of males were in campus leadership positions. These figures have changed greatly over time. Males are still more likely to be in a campus leadership position (41.2% of administratively-certified males remained in the classroom in 2010-11), however 52.89% of administratively-certified females in 2010-11 were either in the assistant or head principal position (leaving 47.11% of females in the classroom).

Tables A.2 through A.8 of the Appendix contain all of the cross tabulations and frequencies calculations for role with ethnicity, role with gender, role across ethnicity, role across gender, role within ethnicity, and role within gender.

### ***Time between significant career events.***

In Chapter three, I proposed to investigate different measures of time in the careers of Texas public school educators with administrative certification, employed at any time between the 1991-92 and 2010-11 school years. I suggested these measures of time would help to understand differences in careers paths of school administrators. These measures mark the passing of time between the occurrences of major events common to the careers of school administrators. Not all events of interest are shared amongst all administrators - for example, not all assistant principals will become head principals. All time variables are measured in years and are defined as follows:

- time2 - time between initial system entry and first administrative certification
- time3 - time between first administrative certification and placement into first campus leadership position (as either assistant principal or head principal).
- time4 - time between first assistant principal assignment and first head principal assignment<sup>13</sup>
- time5 - time between first administrative certification and head principal position
- time6 - time between system entry and any administrative placement

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<sup>13</sup> some individuals jump straight from classroom to head principalship and will not have a value for this measure

time7 - time between system entry and head principal placement

Table 4.1 displays the mean values of times two through seven disaggregated by ethnicity and gender. Because the full population of interest was used to determine these mean values for times two through seven, there is no need to determine the presence of significant difference across groups in table 4.1. In other words, these are the true means and any differences in values between groups are exactly what they are. That being said, these values are compiled only from individuals who successfully made the transition from one role to the next, as defined by the specific time measurement. Therefore, these figures do not take into account any between-group differences in probabilities to procure administrative positions, or the rates at which these procurements occur.

**Table 4.1 - Mean Values of Time by Group**

	Grand mean	Asian	Black	Latino	White	female	male
time2	10.29	7.00	10.22	10.44	10.40	11.36	8.52
time3	14.43	12.63	13.00	13.48	15.11	14.57	14.23
time4	4.07	4.33	4.44	4.07	4.01	4.12	4.00
time5	17.16	15.71	16.91	16.52	17.44	18.12	15.72
time6	13.59	11.91	12.79	13.07	14.01	14.16	12.67
time7	17.16	15.71	16.91	16.52	17.44	18.12	15.72

For the most part, differences between ethnicity groups are fairly small. There are however, exceptions. Asians procure promotion up the administrative hierarchy much more quickly than do Black, Latino, and White educators. Whites take about a year longer to gain administrative certification, and upon doing so, take two additional years to procure a campus leadership position. When it comes to gender, males advance through the system more quickly than do females. On average, males gain their certification three years before females, and make their way to the head principal position about 2.5 years prior to the average female. While certainly informative about educators who have entered the field of public school administration, these measurements of time don't tell us a great deal about the administrator selection process. The following section aims to do just that.

### **Inferential Statistics**

A closer look at how these times differ across various intersections of ethnicity and gender can be constructed through the estimation of survival functions. Survival functions allow the researcher to quantify the rate at which groups of individuals "die" during a particular course of time, while simultaneously accounting for

censored cases. The terms *death* and *failure* in survival functions come from the field of biostatistics, as this type of research was originated to study infant mortality. In this paper, death or failure are not actually bad things, and are defined as the procurement of promotion up the administrative hierarchy.

Let us say that we are interested in how different ethnicity or gender groups experience the transition from initial certification to first assignment as an assistant principal. Procuring an assistant principal job would be considered a *death*. Let us also say that we consider all individuals certified after the 1991-92 school year to be at *hazard* for promotion. Again, hazard is not necessarily a bad thing, but rather indicates that an individual is at risk of experiencing the event of interest - promotion, in this case. Survival functions allow us to see how the paths of varying groups differ over time, while simultaneously taking into consideration individuals who are censored in the data set. To be censored means that an individual did not experience the event at the end of time measurement or that they became ineligible during time measurement. In the case of individuals being promoted to the assistant principalship, an example of a censored individual could be

someone that ascended from the classroom straight to the head principalship - therefore no longer being at hazard to become an assistant principal, or someone who made it to the 2010-11 school year (the final year of the data set) without leaving the classroom.

Survival functions add more nuance to the time measurements explored in the previous section as they allow us to graphically present the rates at which varying groups experience events. The graphical representation can also include intersectionality of ethnicity and gender. First, lets us review a graph representing all study participants.

**Figure 4.7 - Kaplan-Meier - Years to Head  
Principalship - All participants**

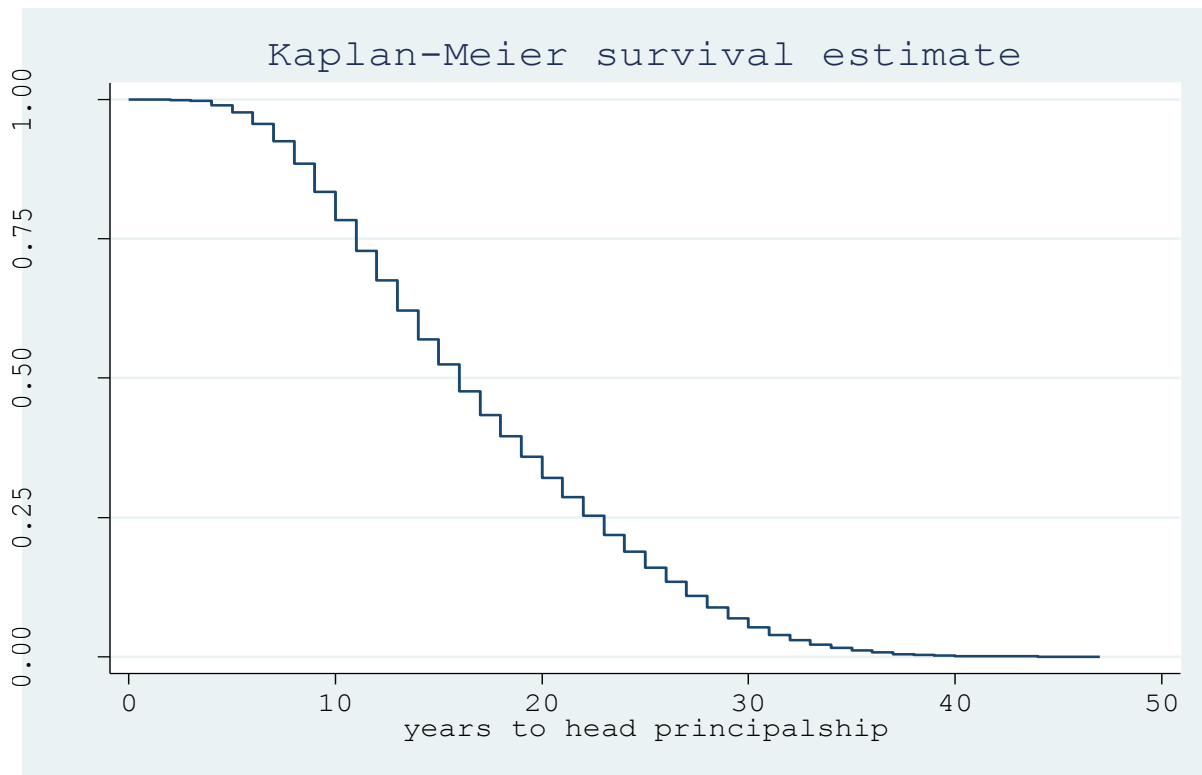


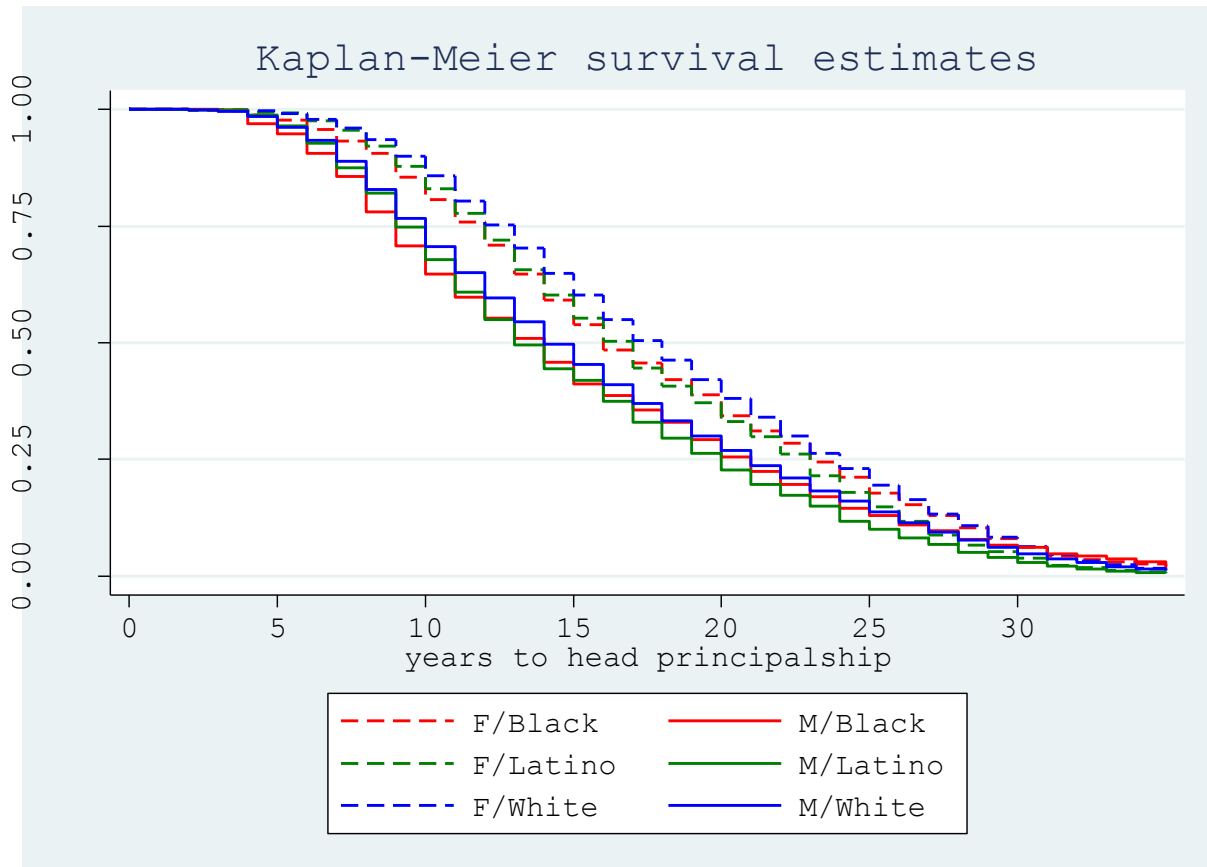
Figure 4.7 displays the hazard rates for all study participants to procure the position of head principalship. The sloping line and its component "steps" represent the change in risk to experience the event of interest (procuring head principalship) as time progresses in years across the x axis. The y axis values represent the proportion of the population still at risk to experience the event at a given time.

Figure 4.8 displays how the survivor functions differ for the three largest ethnicity groups of this study



crossed with gender - each line represents a differing intersection of ethnicity and gender.

**Figure 4.8 - Kaplan-Meier - Years to Head Principalship - Gender by Ethnicity**



As can be seen in the graph, more than half the members of all three male intersectionalities (solid lines) experience the event within 15 years. Black, Latina, and White females (dashed lines) take several more years to get to the same point.

**Figure 4.9 - Kaplan-Meier - Years to Assistant Principalship - Ethnicity by Gender**

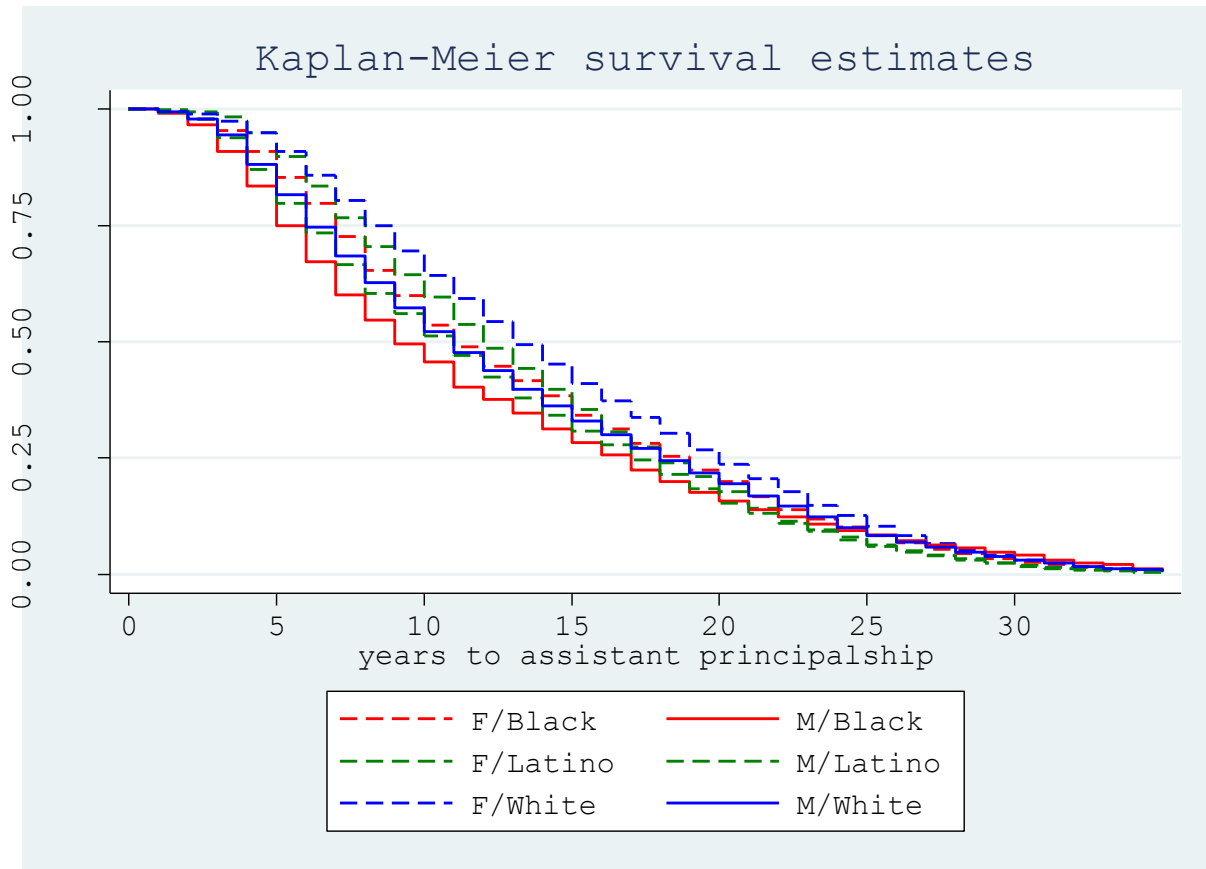
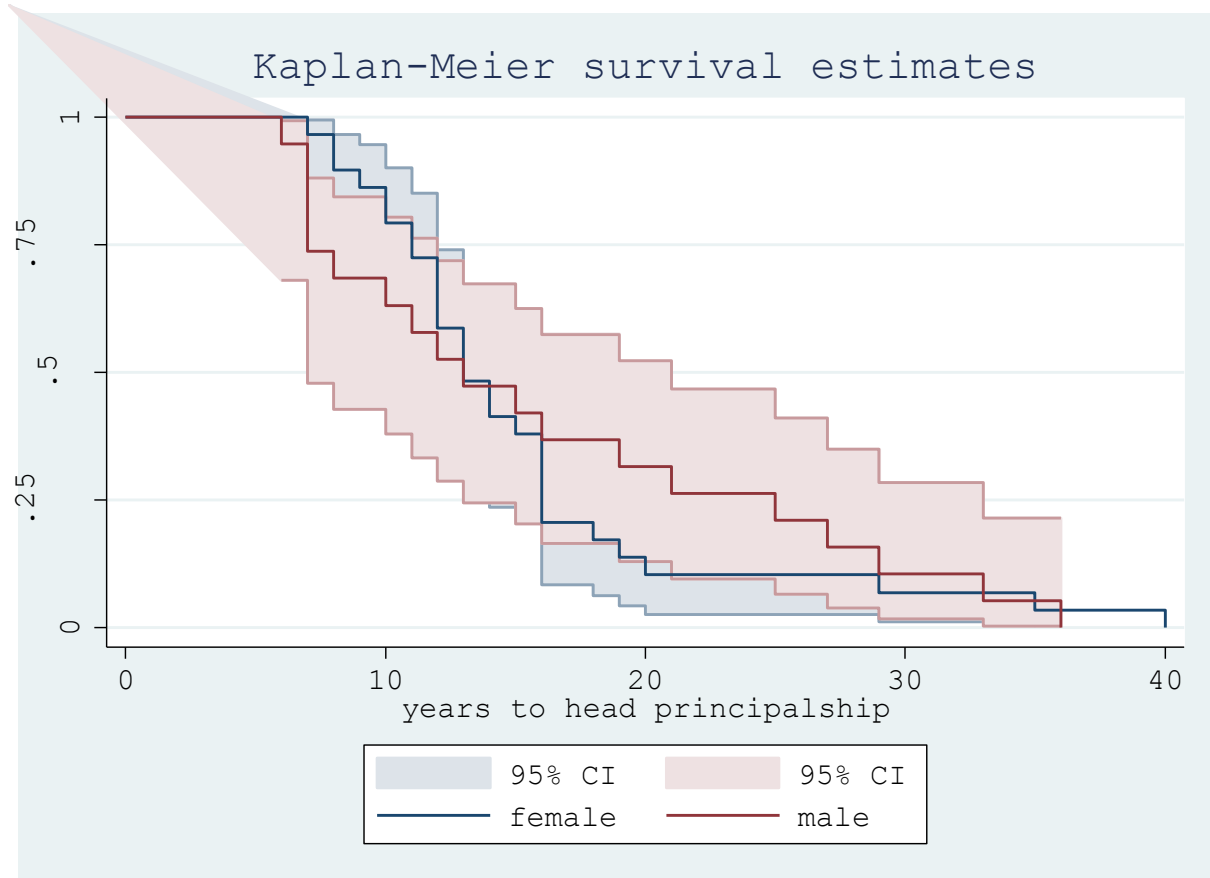


Figure 4.9 displays the same intersectionality groups' survival rates for the assistant principalship position. With the exception of Black males and White females, the separation between intersectionalities is less discernible for the assistant principalship, as evidenced by the tightly packed curves.

It should be noted that because the Asian group for both males and females is so small, the confidence interval for their survival rates is very large. Therefore,

graphical representation of their experiences compared with the other groups is difficult to interpret. The shaded bands surrounding the colored lines for males and females represent the area within which 95% of all cases would be expected to fall. Figure 4.10 is a graphical representation of these confidence intervals. Confidence intervals for all other groups, while not pictured, are very tight

**Figure 4.10 - Kaplan-Meier - Years to Head Principalship - Asians by gender**



In sum, the survival estimates show that disparities between groups are more strongly defined along gender lines than they are for ethnicity.

### **Research question 3.**

The third research question of this paper seeks to determine what factors help explain the disproportionate assignment of campus leaders. Of specific interest is selection bias based upon an individual's ethnicity and gender. In Chapter three I proposed to calculate selection probabilities via multinomial logistic regression.

Multinomial logistic regression is much like binary logistic regression, but instead of calculating how covariates contribute to the odds of achieving one of two discrete outcomes, it allows the researcher to determine how covariates contribute to the odds of achieving one of *several* different outcomes. For this part of the paper, the outcomes of interest are the various roles that administratively-certified educators fill. The "baseline" or normal/expected outcome is to remain in the classroom. The other outcomes are 1) being an assistant principal, and 2) being a head principal. Multinomial logistic regression fixes to a value 0, the log odds of an individual with mean value of all covariates remaining in the classroom. I have

programmed Stata to report relative risk ratios (RRR) for each covariate. These RRRs represent the change in risk an individual would experience with a one unit increase in the covariate of interest. For all levels, the excluded comparison group is White females, since they make the majority of the education workforce<sup>14</sup>. Therefore, all significant RRRs within the results table can be read as the change in an individual's odds to be an assistant or head principal (depending which part of the table the RRR is being taken from), relative to White females, controlling for all other variables.

The table below lists the three levels of the dependent variable "admin". Recall that the base outcome is to remain in the classroom. Level two of admin is being an assistant principal, while level three is being a head principal. The variables listed within each level are the independent variables I have chosen to include in the model as covariates effecting odds for advancement. The  $P > |z|$  column reports the significance of each covariate, while

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<sup>14</sup> While not required, standard practice with multinomial logistic regression is to set the largest subgroup in your data as the comparison group.

the z column reports the relative strength of influence that each covariate has in relation to the others.

Assuming statistical significance, independent variables with RRR values greater than 1.0 mean that a one-unit increase in that particular covariate is positively related to an individual's probability to be an assistant or head principal relative to White females.

Table 4.2 lists five significant covariates ( $p < .05$ ) in the determination of an individual's probability to be promoted: these are being a Black male, being a Latino Male, the main effect of being male, one's age, and one's tenure<sup>15,16</sup>. All of these covariates are related to positive change in an individuals' relative risk with the exception of age. Age has a RRR value of less than 1.0 signifying a negative relationship. A deeper exploration of age and its relation to job procurements is provided in chapter five.

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<sup>15</sup> Table 4.2 is a much simpler model than others in this study created for later school years in which there is more information available. Table 4.2 is testing for the effects of ethnicity and gender, controlling only for an individuals' age and tenure. Age and tenure, while relevant, are hardly an exhaustive selection of independent variables in an ideal model testing for selection bias of administrators.

<sup>16</sup> tenure is defined as an individual's total number of years in the public education system at the year of analysis. It should be noted that

**Table 4.2 - Relative Risk Ratios - 1991-92**

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	.436495	.2342884	-1.54	0.122	.1524395	1.249859
	Black	.8892755	.0699332	-1.49	0.136	.7622497	1.03747
	Latino	1.121119	.081466	1.57	0.116	.9722974	1.292719
	Other	.7036409	.277988	-0.89	0.374	.324389	1.526286
	AsianMale	2.848278	1.787461	1.67	0.095	.8325371	9.744538
	BlackMale	1.816439	.216406	5.01	0.000	1.438172	2.294198
	LatinoMale	1.423529	.1463641	3.43	0.001	1.163718	1.741346
	OtherMale	1.293968	.6524932	0.51	0.609	.4816139	3.476549
	male	1.583787	.0712147	10.23	0.000	1.450183	1.729701
	age	.9702733	.0025921	-11.30	0.000	.9652061	.9753672
	tenure	1.013886	.0026226	5.33	0.000	1.008758	1.019039
3							
	Asian	.7854809	.2962774	-0.64	0.522	.3750325	1.645138
	Black	.7374034	.0586339	-3.83	0.000	.6309903	.8617626
	Latino	1.261752	.0882057	3.33	0.001	1.100193	1.447036
	Other	.3251662	.172459	-2.12	0.034	.1149879	.9195147
	AsianMale	.6710508	.3248825	-0.82	0.410	.2598093	1.73323
	BlackMale	.8493148	.1053615	-1.32	0.188	.6659984	1.083089
	LatinoMale	.7089476	.0707435	-3.45	0.001	.5830095	.8620901
	OtherMale	1.173456	.7461802	0.25	0.801	.3374444	4.080666
	male	2.487052	.1016105	22.30	0.000	2.295664	2.694396
	age	1.008648	.0024744	3.51	0.000	1.00381	1.01351
	tenure	1.032329	.0023519	13.97	0.000	1.027729	1.036949

Results from multinomial logistic regressions are essentially meaningless if the assumption of independence of irrelevant alternatives (IIA) is violated. In this case, as the researcher, I am making the assumption that all alternatives outside of my model - that is to say, the inclusion of a role beyond being a classroom teacher or campus leader as an additional outcome level - would not affect the estimates of my model. If I were to exclude a 4<sup>th</sup> outcome level in which an individual becomes a

superintendent(a level excluded in this paper), I would have to prove that considering such an outcome wouldn't have a significant impact on my model's estimations. For all 20 years in the study, I tested for the independence of irrelevant outcomes in my state-wide models. All models passed the IIA tests with the exception of the 2001-02 model. The results of these tests can be found in Table A.8 of the appendix.

For each of the 20 years of the study, I conducted 20 additional models by region<sup>17</sup>. I used the 20 education service centers designed by the state as regional boundaries. These analyses allow me to determine how factors affecting selection differ throughout various regions of the state and how those differences have changed over time. I did the same for campus types as well. The data set is broken up into elementary, middle, secondary, and alternative campuses. In total, over 400 separate models were estimated to explore factors affecting the selection of school administrators. All of these models can be found in table A.8 of the appendix.

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<sup>17</sup> A map of these regions is provided in Figure A.1 of the Appendix



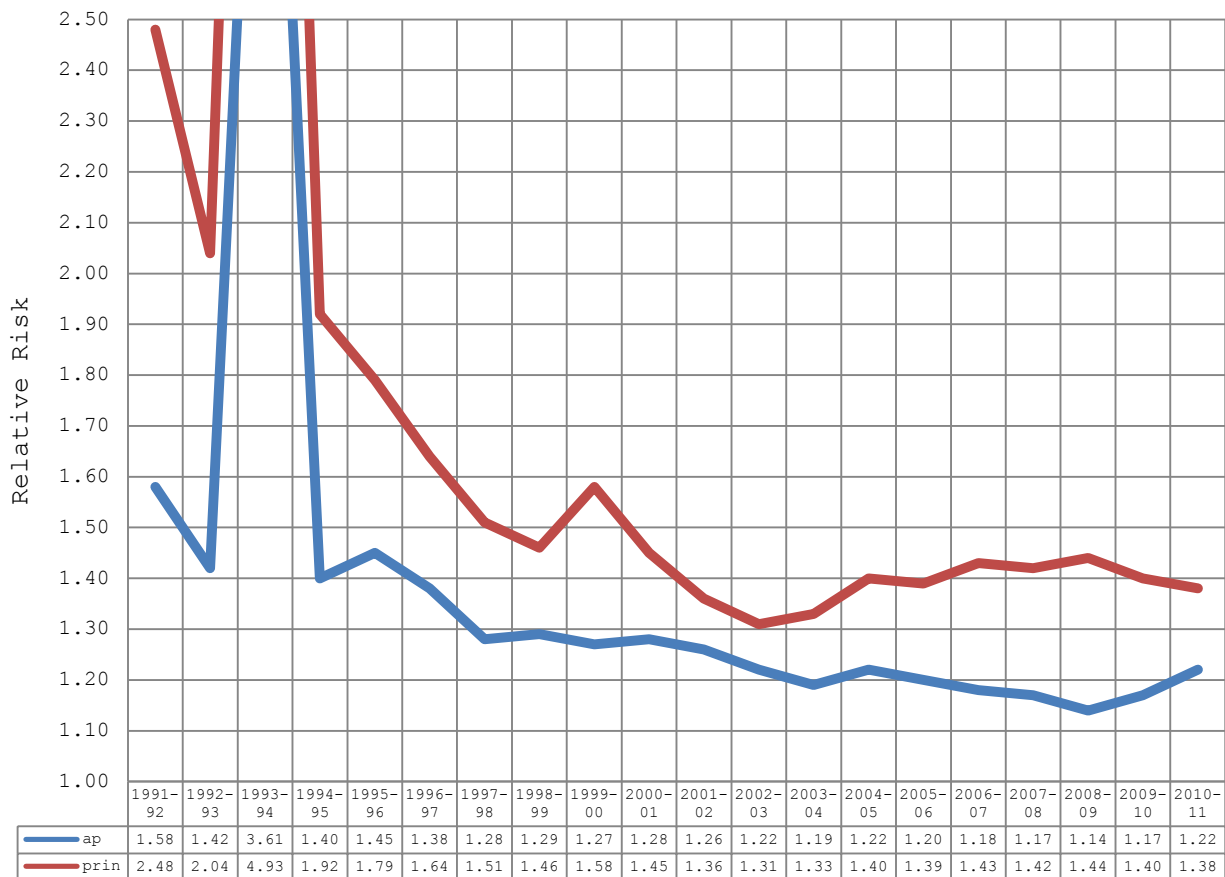
As the years progress from 1991 to 2011, TEA collected an increasingly informative amount of information about its employees. Therefore, the models increase with complexity over time as I'm able to add covariates which more richly describe the campuses in which the certified individuals are employed and the tenure they have built with their current district. Covariates I chose to include along these lines include the percentage of students of limited English proficiency, at risk, economically disadvantaged, Black, White, Latino, and mobile. I also include controls for the accountability rating of the campuses as well. The accountability ratings, determined by TEA and reported via AEIS assign a rating of Academically Unacceptable, Academically Acceptable, Recognized, and Exemplary based on a variety of campus outcomes not limited to, but including student performance on state standardized tests, attendance, dropout rates, and graduation rates.

The most salient themes that came from the 478 multinomial logistic regression models contained in this study are as follows:

- 1) The first models listed under each academic year in table A.8 of the appendix represent state-wide calculations. In all 20 statewide models, the effect of

being male is positively and significantly related to an individual's likelihood to be a campus administrator. The main effect for male is strongest at the middle and secondary school level. Figure 4.11 displays how the state-wide relative risk ratio for the male dependent variable has changed over time. As the years pass, the effect appears to be drawing closer to one, meaning that male advantage in procuring campus leadership positions has waned over time.

**Figure 4.11 - RRR of "Male" over Time**



2) Black males are more likely to be assistant principals than head principals. This holds for virtually all models across time and region. The RRR assigned to the BlackMale covariate is almost always lower in level 3 (head principal) of the output table than in level 2 (assistant principal).

3) Coming from a school with high concentrations of Black, Latino, and mobile students is negatively associated with one's likelihood to become a campus leader.

4) Probabilities vary across school types. White female's strongest probabilities to be campus leaders lie at the elementary level. Non-Whites are more likely to find administrative placement in middle and secondary schools. This is particularly true for non-White males.

5) The RRRs of the independent variables vary across regions. For instance, in the same year, the male main effect in level two of admin varies from insignificant in one region all the way to 3.95 in another.

## **Conclusion**

The analysis chapter of this study has covered a variety of reports and models across the descriptive and inferential statistical spectrums. Beginning with a complete analysis of the study participants, this chapter

segued into a response to research question one, which asked "What is the current state of leadership production in Texas and how has production trended over time?". Descriptive statistics showed a seemingly adequate pipeline of educational administration graduates that has grown steadily over time. The same was found to be true for new administration certificates. Interesting changes in administrator production and selection, to be discussed in the next chapter, were noted beginning in 2009.

Responses to the second research question sought to determine how the careers of administratively-certified educators differ across ethnicity and gender. Results showed that males have been and continue to be overrepresented in school administration. Historical trends have also shown that the White and male domination of school leadership, while still present, has continued to erode over time.

Comparisons of time between promotions of campus leaders, conducted across ethnicity and gender showed significant disparities between males and females. Differences between ethnicities were much smaller. These time comparisons only looked at individuals who had secured said promotions, therefore a closer look at careers by

ethnicity and gender was needed. Survivor functions showed females at a distinct disadvantage in terms of career promotions in school leadership, particularly at the head principal position.

Finally, a large set of multinomial logistic regression models (487) were estimated to see how ethnicity and gender relate to an individual's probability to become an assistant or head principal. A large number of independent variables were entered into these models to control for the overall number of years of experience in the Texas public education system, an individual's tenure as measured by the number of consecutive years employed with the same district, and school setting that the campus leadership candidates were coming from (i.e. accountability rating and student descriptives). These models found serious disparities between ethnicity groups and gender. Most differences in probabilities were explained by interactions between ethnicity and gender where White females were very likely to lead elementary campuses, but were less likely to gain other positions. Males, particularly at the middle and secondary level were far more likely to be campus leaders, especially head principals. The effect of ethnicity on one's probability to

become a campus leader was found to vary greatly across region of the state and school type. In sum, the multinomial logistic regression models created a strong case for the presence of gender and ethnicity-based bias in the selection and promotion of school leaders.

## **Chapter Five: Discussion**

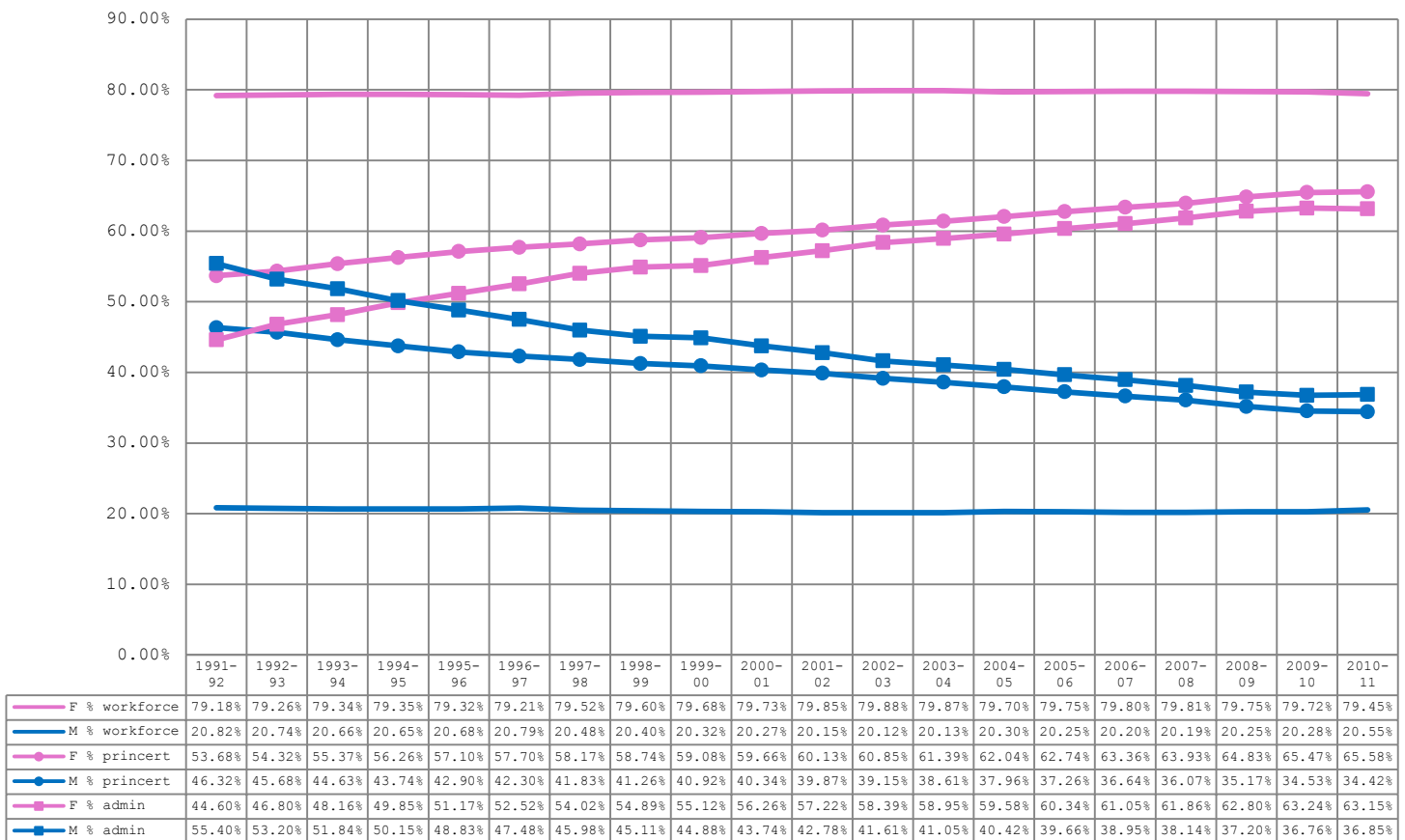
This paper has been an exploration of the overrepresentation of Whites and males amongst Texas public school leadership. Using a variety of descriptive and inferential techniques, I have sought to further understand disparities in the distribution of gender and ethnicity across educative roles for administratively-certified individuals. I have also explored the evidence for bias being amongst the factors affecting administrator selection in various regions, settings, and times across the state. Here in chapter five, I discuss the descriptives and model output from chapter four as I endeavor to tie these data together with the scholarship reviewed in chapter two and my own interpretations of the overrepresentation phenomena.

### **Interpretation of Findings**

Chapter four began with an analysis of all research participants, followed by a compilation of descriptive statistics outlining changes in the representation of campus roles across and within gender over the 20 years of this study. A common thread from all of the descriptive statistics was a decreasing share of Whites and males over time in the state's educator workforce.

As mentioned in chapter four, the gender split over time in the Texas public school workforce has remained static at almost exactly 80% females and 20% males. This is displayed in Figure 5.1 below by the solid pink and blue lines riding the 80% and 20% horizontal gridlines (% workforce).

**Figure 5.1 - Percentage Workforce, Certified Workforce, And Campus Administrators by Gender**



The lines marked with circles represent the shares of females and males amongst Texas public school educators holding valid administrative certification (% princert),

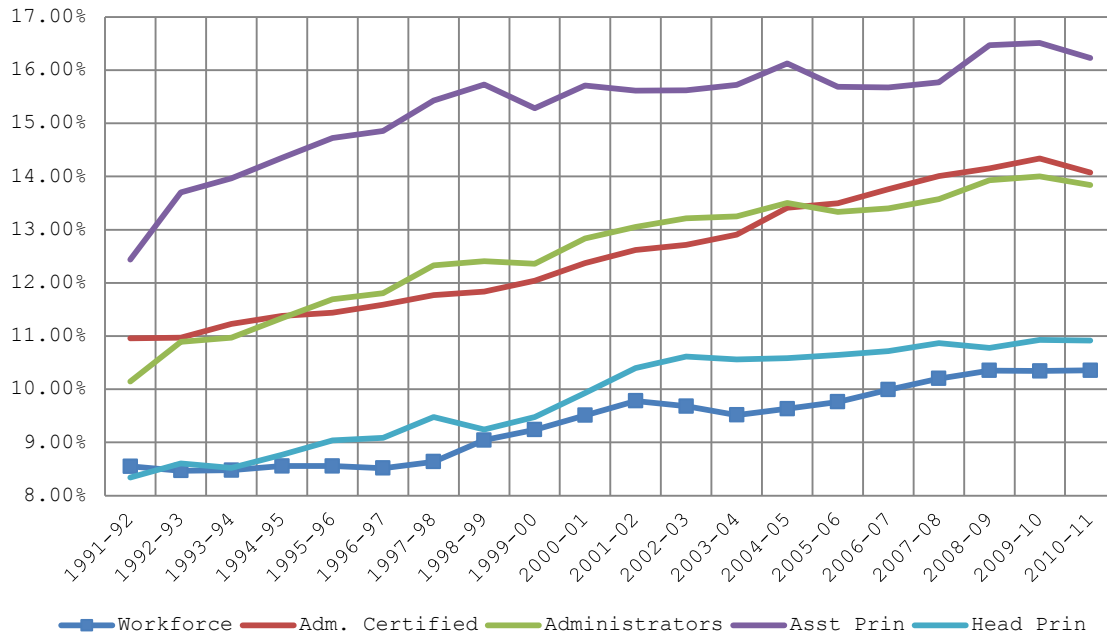


while the squared lines represent the gender breakdown of those assigned to administrative positions (% admin). What is clear from the chart is that for all years in the study, females make a significantly smaller portion of those with administrative certification and campus leadership roles than the overall educator workforce. Despite positive changes over time in the representation of females amongst the administratively-certified and campus leaders, there remains a distinct overrepresentation of males in the present day.

Group representation as it breaks down by ethnicity is a bit more complex, and varies greatly by role. Figures 5.2, 5.3, and 5.4 illustrate the representation of individuals within the Black, Latino, and White ethnicity groups amongst the public school workforce, educators with administrative certification (Adm. Certified), and campus leaders (Administrators). Further still, additional lines and frequencies are added to show how each ethnicity is broken down across the roles of assistant principal and head principal (asst. = assistant principals, prin = head principals). For all three figures, the line with square markers is the percentage of that ethnicity group's

representation amongst the public school educator workforce writ large.

**Figure 5.2 - Black Representation Over Time**



Black public school educators have historically represented a larger share of the administratively-certified and campus leader populations than they have the overall educator workforce. This trend continues today. What is particularly interesting about this is that despite this seeming favorable representation, Black educators have historically held and continue to hold head principal positions at a rate almost identical to their representation amongst the overall educator workforce. In other words, their favorable job placement rates stops at

the head principalship. At the assistant principalship, Black overrepresentation is typically resides at more than 50% of their share of the educator workforce in any given year. These figures suggest willingness amongst decision makers and gatekeepers to progressively support Black educators in school leadership, but only up to the level of assistant principal.

**Figure 5.3 - Latino Representation over Time**

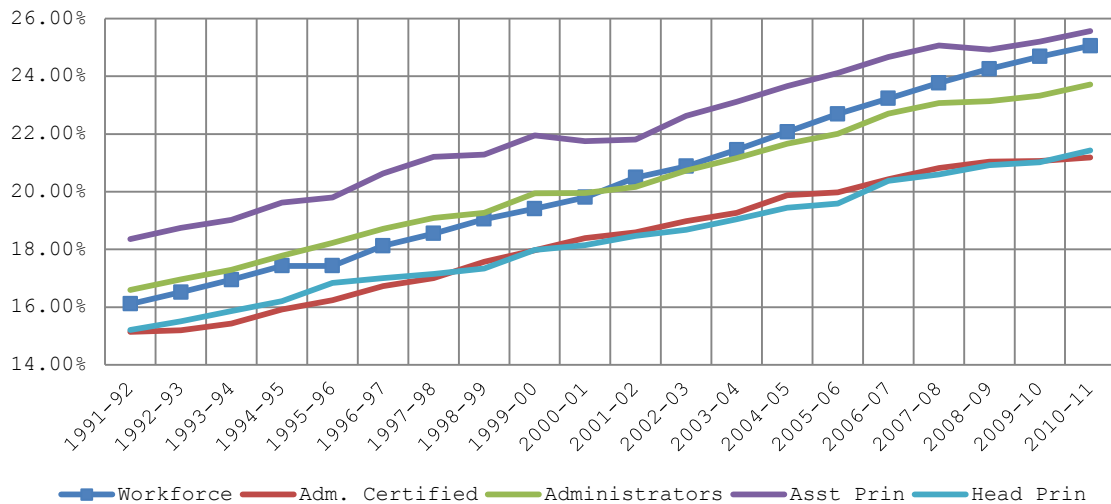
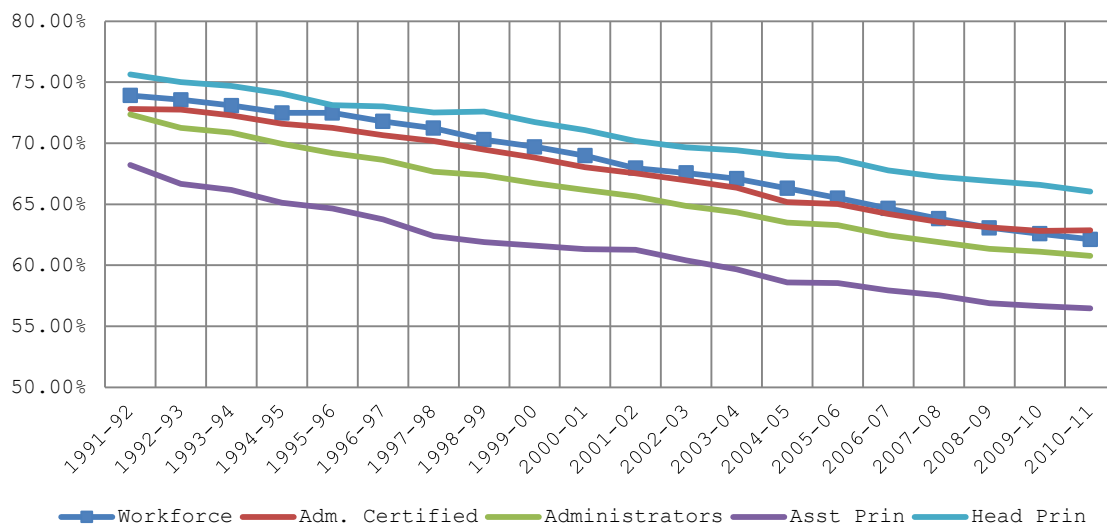


Figure 5.3 shows a tight grouping of employment and certification trends for Latinos in Texas public schools. The only position in which Latinos are overrepresented is in the assistant principalship; however this distinction appears to be shrinking over the past few years. A continuation of the trend lines at their current slopes suggests that Latinos would continue to increase their

share of the public school workforce, but at a rate which would outgrow their representation amongst the administratively-certified and campus leaders. In other words, there is potential for future underrepresentation of Latinos in school leadership across all roles<sup>18</sup>.

**Figure 5.4 – White Representation over Time**



Whites appear to be fairly-equally represented amongst all administrators as they are within the overall educator workforce. Whites, however, have the privilege of having *maintained* and even *expanded* their overrepresentation

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<sup>18</sup> Trends in the representation of Latino educators are of particular concern when results of the 2010 census are brought into consideration. Latinos currently account for 37.6% of Texas' population. One in four Latinos state-wide is of school age (5 to 18). 10% of all Latinos are aged 10 or under (compared to 7.0% of whites and 7.8% of Blacks). Together, these statistics suggests that the Latino share of the state's student and overall populations will continue to grow. (Census Bureau, 2010)

within the head principalship. At the same time, Whites have a large underrepresentation in the assistant principalship. The multinomial logistic regression results had significantly unfavorable RRRs for women to attain assistant leadership positions, particularly at the secondary level where the majority of these jobs exist - this explains the majority of the underrepresentation for Whites in the assistant principalship.

***2009 bumps / Flooded Administrator Market.***

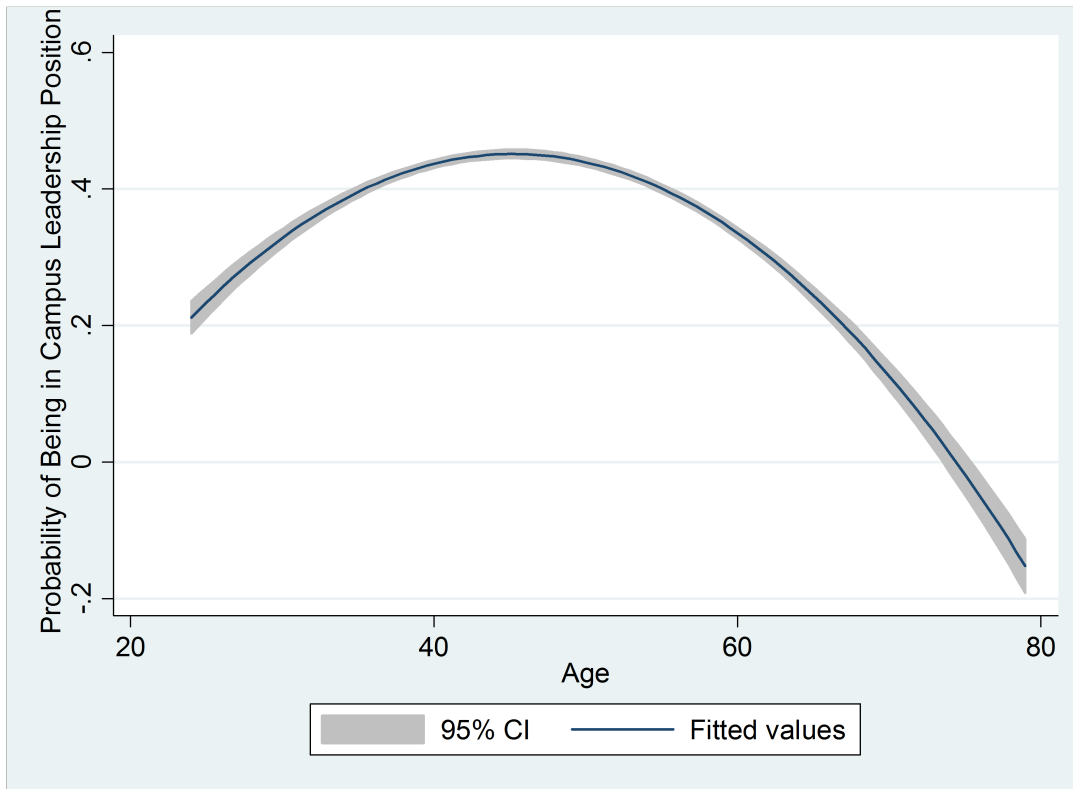
Figures 4.1 and 4.2 have very interesting bumps in the longitudinal trend of educational administration master's degree production and the procurement of first time campus leadership positions. An interesting change occurred in Texas' leadership preparation program landscape in the years preceding these bumps. Lamar University in Beaumont, Texas registered and enrolled its first online cohorts in its new educational administration master's program. The jump in degree production in 2009 is due almost exclusively to Lamar's first cohort graduating. In the Spring of 2009, Lamar had over 3,000 individuals concurrently enrolled in their educational administration master's program - the majority of them White females. The drop off in first time administrator job procurements which occurred in 2009

coincided with Lamar's first cohort's graduation - suggesting that gatekeepers declined to hire graduates of Lamar's program and took their chances with existing principals at a rate much higher than in years past.

***Effect of age on likelihood to be campus leader.***

In many of the multinomial logistic regression models, the effect of age was significant and negative. On the surface, this would suggest that each additional year in age is negatively associated with one's probability of becoming a principal. Surely that is not possible for all educators, especially those new to the profession with relatively new administrative certification. An alternative suggestion would be that there is a non-linear relationship between age and the log odds of procuring an administrative position. Novice educators and those in the twilight of their careers are rarely selected into leadership positions, either because they are not credentialed, aren't applying for those positions, deemed undesirable by gatekeepers, or some combination of these reasons. To test this alternative theory, figure 5.5 was constructed, which represents a quadratic fit between the probability of being an administrator with age.

**Figure 5.5 - Effect of Age**



As can be seen by the graph, the probability of being a campus leader is highest when an educator is in their late 40's. Additional analyses showed the effect of age to be similar across the various intersections of ethnicity and gender.

**Black educators as assistant principals.**

As figure 5.2 illustrated, administratively-certified, Black educators are far more likely to be assistant principals than they are head principals. I suggested this may be due to a propensity for gatekeepers to funnel Black

educators into assistant principal positions while simultaneously filtering them out of head principal positions. Additionally, I suggest this is occurring because district-level gatekeepers and White, male administrators at the middle and secondary level, especially those in urban school settings or campuses where the student population is more likely to contain non-Whites, are hiring Black, male, assistant principals to deal with student discipline.

The disproportionate discipline referral rates for young, Black, males is a crisis in our education system (Johnston, 2000; Monroe, 2005). I posit that Black, male assistant principals are brought in to address this issue in the role of campus disciplinarian.

In their interviews with Black educators about the lasting effects of the *Brown* decision, participants in Milner and Howard's study (2004) explained that Black educators whom were brought over to White schools during integration were put in charge of disciplining Black, male students. The respondents suggested this role assignment for Black administrators continues today, and has dealt a "devastating blow" to the relationship between Black male students and administrators. They explained that "Black



males were forced to see their Black leaders as disciplinarians whom they fear as opposed to Black leaders in whom they could admire and look to for mentorship” (p. 290).

The expectation and assumption for Black male administrators to be disciplinarians is further entrenched via popular media. Gooden (2012) describes how the stories of Joe Clark and Ronald Stone - Black urban school principals popularized in the films *Lean on Me* and *Heart of Stone* respectively - had major influence on the national conscience. These leaders had authoritarian leadership styles and a “no-nonsense” approach to discipline. Gooden explains that as influential individuals such as former secretary of education William Bennett began to champion Clark’s leadership style, the archetypical role of the Black, male, disciplinarian campus administrator was constructed. I believe this archetype remains today and is evidenced in the role assignment of administratively-certified, Black, male educators in Texas.

### **Further Interpretations of Disproportionality**

As I alluded to earlier, and wish to make very clear in this space, I believe that the discrepancies seen in all of the statistical models across gender, ethnicity, and the

various intersectionalities of the two, is due to racism and sexism amongst gatekeepers in our public school system. In our society, the dominant groups, Whites and males respectively, have the distinct privilege of being able to ignore their own race and gender while simultaneously ignoring or discounting the struggles encountered by those unlike themselves. It is this very privilege and the lack of acknowledgement for its existence that further cements the supremacy of Whites and males.

As the literature reviewed in chapter two suggested, people have preconceived notions of the leadership efficacy of individuals formed solely by their outside evaluation of a candidate's ethnicity and gender. I do not believe it is idealistic to assume that an ethnicity or gender group's representation amongst all levels of school leadership should fall in line with their share of the overall educator workforce. In fact, I suggest that to believe otherwise would be an indictment on one's beliefs about the inherent leadership abilities of people based upon their ethnicity and gender.

Much of the racism and sexism in our education system is rooted in our nation's history. In an analysis of New York state teachers employed between 1978 and 1988, Brewer

(1996) found that males were far more likely to procure administrative positions. He also discovered large and significant discrepancies in the salaries of males and females in the same role. Historical discrepancies in pay and responsibility are not unique to New York. These systematic differences in the pay and promotion of public educators act as signals from LEAs and SEAs suggesting that lower value should be placed upon the contributions of women to the education system. When these signals are out there for anyone inside or outside of the education system to see and be influenced by, their lasting effect cannot be ignored.

In another incredibly detailed and thoughtful review of the lasting effects of the *Brown* decision on Black educators, Tillman (2004) makes the argument that the role and influence of Black administrators was greatly damaged during integration. She explains:

During the period immediately following *Brown*, Whites believed that Black children had not learned because Black principals had not been effective in assuring that these children were educated. "Expert witnesses" who testified during a period of constant legal proceedings about the issue of desegregation called for the dismantling of all Black schools and replacing Black principals with White principals. (p. 293)

Further, Tillman cites Yeakey et. al., 1986 who proposed:

Since racial patterns in most communities, especially those in the South did not countenance Blacks supervising Whites in any capacity, much less teaching, principals of formerly Black schools usually were reassigned as assistants to White principals or as central office supervisors. (p. 122)

Together, these quotations illustrate that early-on after the *Brown* decision, the script was flipped on the role and presumed efficacy of Black school leaders. Varying estimates in the article suggest that nearly three out of every four Black public school leaders in the South lost their jobs as a result of *Brown*. There is no universe in which such a widespread dismissal of Black leaders combined with official devaluations of their leadership efficacy could not have a lasting effect on Black school leaders today.

What's at stake when racism and sexism are a problem?

### **Contribution/Significance of Findings**

As a result of this study, we have for the first time, comprehensive, longitudinal information about the trends in gender and ethnicity diversity amongst Texas public school leaders. As the findings have illustrated, we are not experiencing a simple cut and dry instance of White, male domination. Rather, there are stories to be told at the various intersections of ethnicity and gender as well. It

is my hope that this study sounds the alarm for the state and its component districts to revisit their decision-making and staffing strategies.

As I have stated, there is clear evidence of bias in our public education system, the same public education system teaching our kids that they can be anything they want to be, that diversity is a good thing, and that our society is a meritocracy. I have concerns about an education system that fails to mirror its own teachings. What sort of effect does this have on students? They are not blind, they see what is going on with school staffing, and are unquestionably affected by it.

**Policy influence.**

Making available to districts, data similar to that compiled by this study would be a wise policy decision on the part of TEA. Having mindfulness of where you are as a district with regards to staffing diversity, where you've been, and how things are trending over time, can go a long way in making positive changes for the future. I will not pretend that adding an additional layer of accountability to states and districts around their staffing policy is the correct answer. Instead, I think that facts and research have the potential to incentivize change within districts,

regions, and states. In some cases, simply providing awareness of disproportionality can serve as an impetus for change. Others organizations will need further incentive.

One piece of research that may serve an incentivizing role was conducted by Herring (2009), who made the argument that "diversity pays". Herring suggests that organizations' perspectives on diversity fall into one of three categories: 1) the value-in-diversity perspective which believes that a diverse workforce produces better business results, 2) the diversity-as-process-loss perspective which holds that diversity introduces conflict into an organization and negatively effects efficiency, and 3) a paradoxical combination of the first two which believes that diversity introduces group conflict *and* better business performance because the status quo is constantly being challenged. Herring reviewed data from 509 for-profit business organizations and found that higher levels of gender and ethnicity diversity within these organizations were positively related to increases in sales revenues, customer bases, and overall profits. Herring suggested these positive changes come about because increased diversity "allows companies to 'think outside the box' by brining previously excluded groups inside the box ...

enhancing an organization's creativity, problem-solving, and performance" (p. 220).

While school districts and state education systems are not for-profit business organizations, Herring's findings remain encouraging. After all, school districts attempting to duplicate these findings would hardly be the first instance of education mimicking the business world.

### **Blueprint for additional research.**

An additional contribution from this paper is the provision of a blueprint for conducting similar research in areas beyond Texas. I've purposely included a variety of quantitative techniques so that other states, regions, and districts can answer questions about employee diversity that are pertinent to their specific contexts. Additional studies such as these conducted on data from other areas can provide a more rich description of the current state of our nation's educational administrators. I intend to be a point of reference for anyone interested in the methods or coding used in this study.

### **Next Steps**

As with any study, results ask just as many questions as they answer, if not more. One prevailing question I have asks why the main effect for gender has remained so strong

over time. I believe that revisiting my research questions with more nuanced data would provide some additional context in this area. For instance, I did not control for the rate at which individuals *applied to* and accepted *interviews for*, open administrative positions. Administering surveys to a selected subsample of this study's participants could go a long way in determining if there are systematic differences in the rates at which males and females, or even different ethnicities, apply for and accept interviews for campus leadership openings.

An additional data point that would improve future iterations of this research strand would be the inclusion of individual educational history. Adding variables which rate the quality of each educator's undergraduate and graduate institution would be ideal. It can be argued that educational history not only plays a role in the hiring of leadership candidates, but also in the leadership skills said candidates have. One approach to controlling for individual educator's educational history can be found in Fuller, Baker, & Young's (2011) work, which utilized Carnegie classifications to group the leadership preparation programs their study participants came from.



I also would like to broaden my future studies to consider identities beyond the intersectionalities explored in this study. Holvino (2010) calls for studies exploring ethnicity and gender to expand and include measures for class, sexuality, and gender identities beyond the traditional, discrete categories of male and female. She suggests that:

“essentialized identities produced and reproduced through social and organizational practices can be disrupted by the collection and dissemination of ... differentiated stories and narratives that focus on the complexity of identity-subjectivities” (p. 264).

Procuring these additional data is an entirely different discussion, but is certainly possible through the use of well-designed and intelligently-administrated surveys.

Finally, I would like to recognize that research in our discipline excels at pointing out problems while failing to provide solutions. In sincere hopes of avoiding this common pitfall, I find promise in *performative critical diversity studies*. Described by Zanoni, Janssens, Benschop, and Nkomo (2010), performative critical diversity studies explicitly aim to stimulate social change and “take an agentic perspective to show how the concept of diversity can be strategically deployed by practitioners to achieve

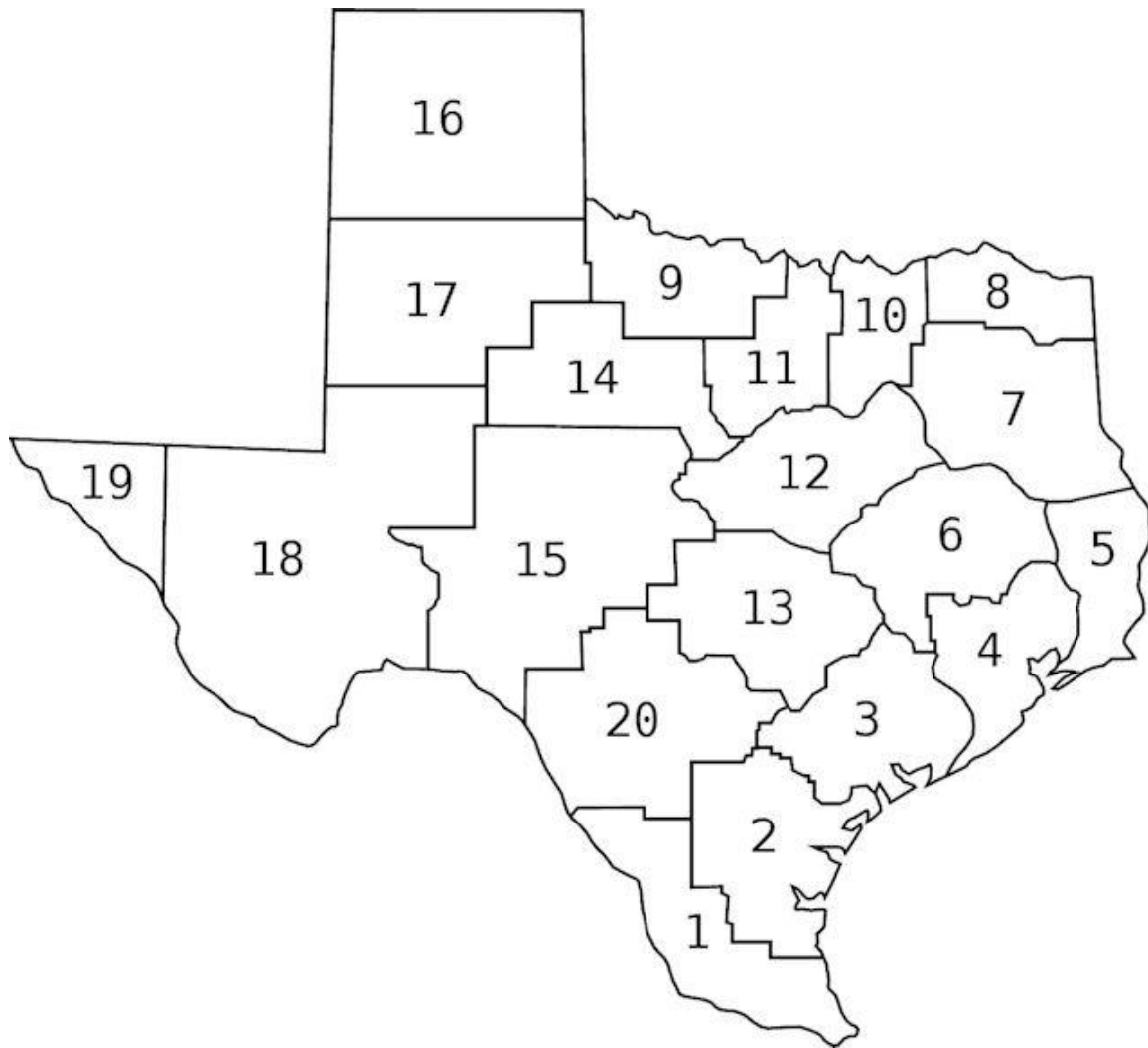
forms of social change" (p. 19). Such studies go beyond participatory research in that strategic, goal-defined, partnerships are created between the researcher and the study participants. I hope that a partnership, or even partnerships of this nature can be formed between myself and organizations that are serious about making positive changes in their administrator selection practice.

### **Concluding Remarks**

Ethnicity and to an extent, even gender, are socially constructed groups we try to force people into - groups even this dissertation has forced people into. I hope it is clear to the reader why such classifications were used for this study. I am working toward a progressive cause in an attempt to vanquish the status quo. It is my sincerest hope that this paper has created a critical space to discuss underrepresentation in Texas public school administration so that positive, permanent change can come about in the near future.

## Appendix

Figure A.1 - Map of Regional Education Service Centers



**Table A.1 - Crosstabs of Gender and Ethnicity**

1991-92	Ethnicity	Female	Male	Total
	Asian	1,284	421	1,705
	Black	19,586	4,341	23,927
	Latino	34,959	10,128	45,087
	Other	1,816	437	2,253
	White	163,911	42,914	206,825
	Total	221,556	58,241	279,797
1992-93	Ethnicity	Female	Male	Total
	Asian	1,330	444	1,774
	Black	20,156	4,552	24,708
	Latino	37,291	10,901	48,192
	Other	1,982	489	2,471
	White	170,455	44,127	214,582
	Total	231,214	60,513	291,727
1993-94	Ethnicity	Female	Male	Total
	Asian	1,419	452	1,871
	Black	20,757	4,806	25,563
	Latino	39,567	11,535	51,102
	Other	2,112	517	2,629
	White	175,377	44,978	220,355
	Total	239,232	62,288	301,520
1994-95	Ethnicity	Female	Male	Total
	Asian	1,522	479	2,001
	Black	21,826	5,119	26,945
	Latino	42,554	12,338	54,892
	Other	2,241	576	2,817
	White	181,765	46,516	228,281
	Total	249,908	65,028	314,936
1995-96	Ethnicity	Female	Male	Total
	Asian	1,618	475	2,093

	Black	22,356	5,273	27,629
	Latino	44,575	12,974	57,549
	Other	2,324	625	2,949
	White	185,500	47,512	233,012
	Total	256,373	66,859	323,232
1996-97	Ethnicity	Female	Male	Total
	Asian	1,674	487	2,161
	Black	22,732	5,553	28,285
	Latino	46,476	13,716	60,192
	Other	2,384	650	3,034
	White	189,756	48,628	238,384
	Total	263,022	69,034	332,056
1997-98	Ethnicity	Female	Male	Total
	Asian	1,752	493	2,245
	Black	24,065	5,824	29,889
	Latino	49,862	14,348	64,210
	Other	2,527	686	3,213
	White	197,010	49,524	246,534
	Total	275,216	70,875	346,091
1998-99	Ethnicity	Female	Male	Total
	Asian	1,900	516	2,416
	Black	26,052	6,296	32,348
	Latino	53,121	15,016	68,137
	Other	2,604	732	3,336
	White	201,031	50,410	251,441
	Total	284,708	72,970	357,678
1999-00	Ethnicity	Female	Male	Total
	Asian	1,991	555	2,546
	Black	27,437	6,613	34,050
	Latino	55,995	15,554	71,549
	Other	2,728	787	3,515
	White	205,528	51,378	256,906
	Total	293,679	74,887	368,566
2000-01	Ethnicity	Female	Male	Total
	Asian	2,140	591	2,731
	Black	29,007	7,036	36,043
	Latino	58,868	16,173	75,041

	Other	2,845	866	3,711
	White	209,277	52,155	261,432
	Total	302,137	76,821	378,958
2001-02	Ethnicity	Female	Male	Total
	Asian	2,370	604	2,974
	Black	30,891	7,460	38,351
	Latino	63,224	17,121	80,345
	Other	3,017	946	3,963
	White	213,615	52,886	266,501
	Total	313,117	79,017	392,134
2002-03	Ethnicity	Female	Male	Total
	Asian	2,688	692	3,380
	Black	30,968	7,624	38,592
	Latino	65,557	17,664	83,221
	Other	3,119	969	4,088
	White	216,113	53,240	269,353
	Total	318,445	80,189	398,634
2003-04	Ethnicity	Female	Male	Total
	Asian	2,909	740	3,649
	Black	30,591	7,467	38,058
	Latino	67,496	18,281	85,777
	Other	3,101	997	4,098
	White	215,299	53,027	268,326
	Total	319,396	80,512	399,908
2004-05	Ethnicity	Female	Male	Total
	Asian	3,088	767	3,855
	Black	31,340	7,859	39,199
	Latino	70,435	19,386	89,821
	Other	3,140	1,070	4,210
	White	216,350	53,521	269,871
	Total	324,353	82,603	406,956
2005-06	Ethnicity	Female	Male	Total
	Asian	3,274	812	4,086
	Black	32,591	8,168	40,759
	Latino	74,282	20,474	94,756
	Other	3,300	1,136	4,436
	White	219,551	53,979	273,530

	Total	332,998	84,569	417,567
2006-07	Ethnicity	Female	Male	Total
	Asian	3,520	885	4,405
	Black	34,250	8,667	42,917
	Latino	78,250	21,514	99,764
	Other	3,508	1,180	4,688
	White	223,198	54,527	277,725
	Total	342,726	86,773	429,499
2007-08	Ethnicity	Female	Male	Total
	Asian	3,817	935	4,752
	Black	36,044	9,181	45,225
	Latino	82,739	22,601	105,340
	Other	3,808	1,280	5,088
	White	227,369	55,498	282,867
	Total	353,777	89,495	443,272
2008-09	Ethnicity	Female	Male	Total
	Asian	4,001	1,017	5,018
	Black	37,099	9,570	46,669
	Latino	85,886	23,473	109,359
	Other	4,075	1,362	5,437
	White	228,484	55,885	284,369
	Total	359,545	91,307	450,852
2009-10	Ethnicity	Female	Male	Total
	Asian	4,261	1,069	5,330
	Black	37,744	9,973	47,717
	Latino	89,607	24,262	113,869
	Other	4,296	1,419	5,715
	White	231,878	56,866	288,744
	Total	367,786	93,589	461,375
2010-11	Ethnicity	Female	Male	Total
	Asian	4,382	1,123	5,505
	Black	37,592	10,178	47,770
	Latino	90,682	24,893	115,575
	Other	4,415	1,473	5,888
	White	229,435	57,132	286,567
	Total	366,506	94,799	461,305

**Table A.2 - Cross Tabulations of Role with Ethnicity**

1991-92		Asian	Black	Latino	Other	White	Total
	ap	19	527	778	23	2,890	4,237
	prin	29	448	817	15	4,063	5,372
	teacher	55	1,149	1,342	69	7,163	9,778
	Total	103	2,124	2,937	107	14,116	19,387
1992-93		Asian	Black	Latino	Other	White	Total
	ap	21	600	821	17	2,920	4,379
	prin	26	462	833	21	4,028	5,370
	teacher	53	1,119	1,367	74	7,516	10,129
	Total	100	2,181	3,021	112	14,464	19,878
1993-94		Asian	Black	Latino	Other	White	Total
	ap	18	626	853	20	2,966	4,483
	prin	25	467	870	25	4,095	5,482
	teacher	47	1,190	1,415	77	7,636	10,365
	Total	90	2,283	3,138	122	14,697	20,330
1994-95		Asian	Black	Latino	Other	White	Total
	ap	13	672	919	29	3,049	4,682
	prin	26	483	893	27	4,079	5,508
	teacher	43	1,231	1,526	89	7,884	10,773
	Total	82	2,386	3,338	145	15,012	20,963
1995-96		Asian	Black	Latino	Other	White	Total
	ap	16	722	971	24	3,171	4,904
	prin	25	507	945	31	4,102	5,610
	teacher	46	1,210	1,547	85	7,915	10,803
	Total	87	2,439	3,463	140	15,188	21,317
1996-97		Asian	Black	Latino	Other	White	Total
	ap	15	761	1,057	24	3,266	5,123
	prin	24	521	975	27	4,187	5,734
	teacher	47	1,237	1,604	88	7,902	10,878
	Total	86	2,519	3,636	139	15,355	21,735
1997-98		Asian	Black	Latino	Other	White	Total
	ap	20	836	1,149	33	3,381	5,419
	prin	20	557	1,008	31	4,262	5,878
	teacher	41	1,222	1,621	89	7,947	10,920
	Total	81	2,615	3,778	153	15,590	22,217
1998-99		Asian	Black	Latino	Other	White	Total
	ap	25	900	1,218	37	3,541	5,721



	prin	18	556	1,043	32	4,366	6,015
	teacher	48	1,273	1,790	100	8,112	11,323
	Total	91	2,729	4,051	169	16,019	23,059
1999-00		Asian	Black	Latino	Other	White	Total
	ap	27	922	1,324	43	3,717	6,033
	prin	17	580	1,100	32	4,389	6,118
	teacher	56	1,318	1,784	101	8,006	11,265
	Total	100	2,820	4,208	176	16,112	23,416
2000-01		Asian	Black	Latino	Other	White	Total
	ap	30	994	1,376	47	3,880	6,327
	prin	16	621	1,135	37	4,445	6,254
	teacher	59	1,296	1,816	98	7,682	10,951
	Total	105	2,911	4,327	182	16,007	23,532
2001-02		Asian	Black	Latino	Other	White	Total
	ap	33	1,023	1,429	52	4,015	6,552
	prin	19	657	1,167	40	4,435	6,318
	teacher	51	1,368	1,896	109	7,865	11,289
	Total	103	3,048	4,492	201	16,315	24,159
2002-03		Asian	Black	Latino	Other	White	Total
	ap	33	1,070	1,550	60	4,136	6,849
	prin	23	674	1,186	43	4,422	6,348
	teacher	56	1,388	1,938	119	7,934	11,435
	Total	112	3,132	4,674	222	16,492	24,632
2003-04		Asian	Black	Latino	Other	White	Total
	ap	39	1,098	1,614	66	4,167	6,984
	prin	20	680	1,226	42	4,469	6,437
	teacher	76	1,503	2,059	131	8,230	11,999
	Total	135	3,281	4,899	239	16,866	25,420
2004-05		Asian	Black	Latino	Other	White	Total
	ap	45	1,178	1,728	74	4,280	7,305
	prin	18	697	1,280	50	4,539	6,584
	teacher	69	1,608	2,153	143	8,101	12,074
	Total	132	3,483	5,161	267	16,920	25,963
2005-06		Asian	Black	Latino	Other	White	Total
	ap	46	1,194	1,835	80	4,456	7,611
	prin	22	708	1,303	48	4,569	6,650
	teacher	67	1,643	2,111	132	8,055	12,008
	Total	135	3,545	5,249	260	17,080	26,269
2006-		Asian	Black	Latino	Other	White	Total

07	ap	48	1,256	1,977	90	4,643	8,014
	prin	28	728	1,384	49	4,602	6,791
	teacher	78	1,736	2,164	141	8,109	12,228
	Total	154	3,720	5,525	280	17,354	27,033
2007-08		Asian	Black	Latino	Other	White	Total
	ap	44	1,336	2,124	92	4,876	8,472
	prin	31	748	1,418	58	4,628	6,883
	teacher	86	1,806	2,240	140	8,147	12,419
	Total	161	3,890	5,782	290	17,651	27,774
2008-09		Asian	Black	Latino	Other	White	Total
	ap	54	1,434	2,170	96	4,953	8,707
	prin	32	757	1,469	67	4,698	7,023
	teacher	89	1,900	2,444	154	8,586	13,173
	Total	175	4,091	6,083	317	18,237	28,903
2009-10		Asian	Black	Latino	Other	White	Total
	ap	56	1,456	2,222	89	4,995	8,818
	prin	32	785	1,510	73	4,784	7,184
	teacher	117	2,085	2,623	173	9,171	14,169
	Total	205	4,326	6,355	335	18,950	30,171
2010-11		Asian	Black	Latino	Other	White	Total
	ap	66	1,445	2,277	88	5,030	8,906
	prin	33	793	1,557	84	4,797	7,264
	teacher	102	1,906	2,402	179	8,681	13,270
	Total	201	4,144	6,236	351	18,508	29,440

**Table A.3 - Frequencies of Role across Ethnicity**

		role representation across race				
1991-92		Asian	Black	Latino	Other	White
	ap	0.45%	12.44%	18.36%	0.54%	68.21%
	prin	0.54%	8.34%	15.21%	0.28%	75.63%
	teacher	0.56%	11.75%	13.72%	0.71%	73.26%
	Total	0.53%	10.96%	15.15%	0.55%	72.81%
1992-93		Asian	Black	Latino	Other	White
	ap	0.48%	13.70%	18.75%	0.39%	66.68%
	prin	0.48%	8.60%	15.51%	0.39%	75.01%
	teacher	0.52%	11.05%	13.50%	0.73%	74.20%
	Total	0.50%	10.97%	15.20%	0.56%	72.76%
1993-94		Asian	Black	Latino	Other	White
	ap	0.40%	13.96%	19.03%	0.45%	66.16%
	prin	0.46%	8.52%	15.87%	0.46%	74.70%
	teacher	0.45%	11.48%	13.65%	0.74%	73.67%
	Total	0.44%	11.23%	15.44%	0.60%	72.29%
1994-95		Asian	Black	Latino	Other	White
	ap	0.28%	14.35%	19.63%	0.62%	65.12%
	prin	0.47%	8.77%	16.21%	0.49%	74.06%
	teacher	0.40%	11.43%	14.17%	0.83%	73.18%
	Total	0.39%	11.38%	15.92%	0.69%	71.61%
1995-96		Asian	Black	Latino	Other	White
	ap	0.33%	14.72%	19.80%	0.49%	64.66%
	prin	0.45%	9.04%	16.84%	0.55%	73.12%
	teacher	0.43%	11.20%	14.32%	0.79%	73.27%
	Total	0.41%	11.44%	16.25%	0.66%	71.25%
1996-97		Asian	Black	Latino	Other	White
	ap	0.29%	14.85%	20.63%	0.47%	63.75%
	prin	0.42%	9.09%	17.00%	0.47%	73.02%
	teacher	0.43%	11.37%	14.75%	0.81%	72.64%
	Total	0.40%	11.59%	16.73%	0.64%	70.65%
1997-98		Asian	Black	Latino	Other	White
	ap	0.37%	15.43%	21.20%	0.61%	62.39%
	prin	0.34%	9.48%	17.15%	0.53%	72.51%
	teacher	0.38%	11.19%	14.84%	0.82%	72.77%
	Total	0.36%	11.77%	17.00%	0.69%	70.17%
1998-		Asian	Black	Latino	Other	White

99	ap	0.44%	15.73%	21.29%	0.65%	61.89%
	prin	0.30%	9.24%	17.34%	0.53%	72.59%
	teacher	0.42%	11.24%	15.81%	0.88%	71.64%
	Total	0.39%	11.83%	17.57%	0.73%	69.47%
1999-00		Asian	Black	Latino	Other	White
	ap	0.45%	15.28%	21.95%	0.71%	61.61%
	prin	0.28%	9.48%	17.98%	0.52%	71.74%
	teacher	0.50%	11.70%	15.84%	0.90%	71.07%
	Total	0.43%	12.04%	17.97%	0.75%	68.81%
2000-01		Asian	Black	Latino	Other	White
	ap	0.47%	15.71%	21.75%	0.74%	61.32%
	prin	0.26%	9.93%	18.15%	0.59%	71.07%
	teacher	0.54%	11.83%	16.58%	0.89%	70.15%
	Total	0.45%	12.37%	18.39%	0.77%	68.02%
2001-02		Asian	Black	Latino	Other	White
	ap	0.50%	15.61%	21.81%	0.79%	61.28%
	prin	0.30%	10.40%	18.47%	0.63%	70.20%
	teacher	0.45%	12.12%	16.80%	0.97%	69.67%
	Total	0.43%	12.62%	18.59%	0.83%	67.53%
2002-03		Asian	Black	Latino	Other	White
	ap	0.48%	15.62%	22.63%	0.88%	60.39%
	prin	0.36%	10.62%	18.68%	0.68%	69.66%
	teacher	0.49%	12.14%	16.95%	1.04%	69.38%
	Total	0.45%	12.72%	18.98%	0.90%	66.95%
2003-04		Asian	Black	Latino	Other	White
	ap	0.56%	15.72%	23.11%	0.95%	59.66%
	prin	0.31%	10.56%	19.05%	0.65%	69.43%
	teacher	0.63%	12.53%	17.16%	1.09%	68.59%
	Total	0.53%	12.91%	19.27%	0.94%	66.35%
2004-05		Asian	Black	Latino	Other	White
	ap	0.62%	16.13%	23.66%	1.01%	58.59%
	prin	0.27%	10.59%	19.44%	0.76%	68.94%
	teacher	0.57%	13.32%	17.83%	1.18%	67.09%
	Total	0.51%	13.42%	19.88%	1.03%	65.17%
2005-06		Asian	Black	Latino	Other	White
	ap	0.60%	15.69%	24.11%	1.05%	58.55%
	prin	0.33%	10.65%	19.59%	0.72%	68.71%
	teacher	0.56%	13.68%	17.58%	1.10%	67.08%
	Total	0.51%	13.49%	19.98%	0.99%	65.02%

2006-07		Asian	Black	Latino	Other	White
	ap	0.60%	15.67%	24.67%	1.12%	57.94%
	prin	0.41%	10.72%	20.38%	0.72%	67.77%
	teacher	0.64%	14.20%	17.70%	1.15%	66.32%
	Total	0.57%	13.76%	20.44%	1.04%	64.20%
2007-08		Asian	Black	Latino	Other	White
	ap	0.52%	15.77%	25.07%	1.09%	57.55%
	prin	0.45%	10.87%	20.60%	0.84%	67.24%
	teacher	0.69%	14.54%	18.04%	1.13%	65.60%
	Total	0.58%	14.01%	20.82%	1.04%	63.55%
2008-09		Asian	Black	Latino	Other	White
	ap	0.62%	16.47%	24.92%	1.10%	56.89%
	prin	0.46%	10.78%	20.92%	0.95%	66.89%
	teacher	0.68%	14.42%	18.55%	1.17%	65.18%
	Total	0.61%	14.15%	21.05%	1.10%	63.10%
2009-10		Asian	Black	Latino	Other	White
	ap	0.64%	16.51%	25.20%	1.01%	56.65%
	prin	0.45%	10.93%	21.02%	1.02%	66.59%
	teacher	0.83%	14.72%	18.51%	1.22%	64.73%
	Total	0.68%	14.34%	21.06%	1.11%	62.81%
2010-11		Asian	Black	Latino	Other	White
	ap	0.74%	16.23%	25.57%	0.99%	56.48%
	prin	0.45%	10.92%	21.43%	1.16%	66.04%
	teacher	0.77%	14.36%	18.10%	1.35%	65.42%
	Total	0.68%	14.08%	21.18%	1.19%	62.87%

**Table A.4 - Frequencies of Role within Ethnicity**

	role representation within race					
1991-92		Asian	Black	Latino	Other	White
	ap	18.45%	24.81%	26.49%	21.50%	20.47%
	prin	28.16%	21.09%	27.82%	14.02%	28.78%
	teacher	53.40%	54.10%	45.69%	64.49%	50.74%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1992-93		Asian	Black	Latino	Other	White
	ap	21.00%	27.51%	27.18%	15.18%	20.19%
	prin	26.00%	21.18%	27.57%	18.75%	27.85%
	teacher	53.00%	51.31%	45.25%	66.07%	51.96%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1993-94		Asian	Black	Latino	Other	White
	ap	20.00%	27.42%	27.18%	16.39%	20.18%
	prin	27.78%	20.46%	27.72%	20.49%	27.86%
	teacher	52.22%	52.12%	45.09%	63.11%	51.96%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1994-95		Asian	Black	Latino	Other	White
	ap	15.85%	28.16%	27.53%	20.00%	20.31%
	prin	31.71%	20.24%	26.75%	18.62%	27.17%
	teacher	52.44%	51.59%	45.72%	61.38%	52.52%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1995-96		Asian	Black	Latino	Other	White
	ap	18.39%	29.60%	28.04%	17.14%	20.88%
	prin	28.74%	20.79%	27.29%	22.14%	27.01%
	teacher	52.87%	49.61%	44.67%	60.71%	52.11%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1996-97		Asian	Black	Latino	Other	White
	ap	17.44%	30.21%	29.07%	17.27%	21.27%
	prin	27.91%	20.68%	26.82%	19.42%	27.27%
	teacher	54.65%	49.11%	44.11%	63.31%	51.46%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1997-98		Asian	Black	Latino	Other	White
	ap	24.69%	31.97%	30.41%	21.57%	21.69%
	prin	24.69%	21.30%	26.68%	20.26%	27.34%
	teacher	50.62%	46.73%	42.91%	58.17%	50.97%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1998-		Asian	Black	Latino	Other	White

99	ap	27.47%	32.98%	30.07%	21.89%	22.11%
	prin	19.78%	20.37%	25.75%	18.93%	27.26%
	teacher	52.75%	46.65%	44.19%	59.17%	50.64%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
1999-00		Asian	Black	Latino	Other	White
	ap	27.00%	32.70%	31.46%	24.43%	23.07%
	prin	17.00%	20.57%	26.14%	18.18%	27.24%
	teacher	56.00%	46.74%	42.40%	57.39%	49.69%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2000-01		Asian	Black	Latino	Other	White
	ap	28.57%	34.15%	31.80%	25.82%	24.24%
	prin	15.24%	21.33%	26.23%	20.33%	27.77%
	teacher	56.19%	44.52%	41.97%	53.85%	47.99%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2001-02		Asian	Black	Latino	Other	White
	ap	32.04%	33.56%	31.81%	25.87%	24.61%
	prin	18.45%	21.56%	25.98%	19.90%	27.18%
	teacher	49.51%	44.88%	42.21%	54.23%	48.21%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2002-03		Asian	Black	Latino	Other	White
	ap	29.46%	34.16%	33.16%	27.03%	25.08%
	prin	20.54%	21.52%	25.37%	19.37%	26.81%
	teacher	50.00%	44.32%	41.46%	53.60%	48.11%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2003-04		Asian	Black	Latino	Other	White
	ap	28.89%	33.47%	32.95%	27.62%	24.71%
	prin	14.81%	20.73%	25.03%	17.57%	26.50%
	teacher	56.30%	45.81%	42.03%	54.81%	48.80%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2004-05		Asian	Black	Latino	Other	White
	ap	34.09%	33.82%	33.48%	27.72%	25.30%
	prin	13.64%	20.01%	24.80%	18.73%	26.83%
	teacher	52.27%	46.17%	41.72%	53.56%	47.88%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2005-06		Asian	Black	Latino	Other	White
	ap	34.07%	33.68%	34.96%	30.77%	26.09%
	prin	16.30%	19.97%	24.82%	18.46%	26.75%
	teacher	49.63%	46.35%	40.22%	50.77%	47.16%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%

2006-07		Asian	Black	Latino	Other	White
	ap	31.17%	33.76%	35.78%	32.14%	26.75%
	prin	18.18%	19.57%	25.05%	17.50%	26.52%
	teacher	50.65%	46.67%	39.17%	50.36%	46.73%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2007-08		Asian	Black	Latino	Other	White
	ap	27.33%	34.34%	36.73%	31.72%	27.62%
	prin	19.25%	19.23%	24.52%	20.00%	26.22%
	teacher	53.42%	46.43%	38.74%	48.28%	46.16%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2008-09		Asian	Black	Latino	Other	White
	ap	30.86%	35.05%	35.67%	30.28%	27.16%
	prin	18.29%	18.50%	24.15%	21.14%	25.76%
	teacher	50.86%	46.44%	40.18%	48.58%	47.08%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2009-10		Asian	Black	Latino	Other	White
	ap	27.32%	33.66%	34.96%	26.57%	26.36%
	prin	15.61%	18.15%	23.76%	21.79%	25.25%
	teacher	57.07%	48.20%	41.27%	51.64%	48.40%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%
2010-11		Asian	Black	Latino	Other	White
	ap	32.84%	34.87%	36.51%	25.07%	27.18%
	prin	16.42%	19.14%	24.97%	23.93%	25.92%
	teacher	50.75%	45.99%	38.52%	51.00%	46.90%
	Total	100.00%	100.00%	100.00%	100.00%	100.00%



**Table A.5 - Cross Tabulations of Role with Gender**

1991-92		Female	Male	Total
	ap	2,064	2,171	4,235
	prin	2,220	3,151	5,371
	teacher	6,119	3,655	9,774
	Total	10,403	8,977	19,380
1992-93		Female	Male	
	ap	2,200	2,178	4,378
	prin	2,361	3,007	5,368
	teacher	6,232	3,893	10,125
	Total	10,793	9,078	19,871
1993-94		Female	Male	
	ap	2,266	2,216	4,482
	prin	2,532	2,948	5,480
	teacher	6,454	3,906	10,360
	Total	11,252	9,070	20,322
1994-95		Female	Male	
	ap	2,425	2,256	4,681
	prin	2,654	2,853	5,507
	teacher	6,712	4,058	10,770
	Total	11,791	9,167	20,958
1995-96		Female	Male	
	ap	2,597	2,306	4,903
	prin	2,782	2,827	5,609
	teacher	6,791	4,010	10,801
	Total	12,170	9,143	21,313
1996-97		Female	Male	
	ap	2,767	2,355	5,122
	prin	2,934	2,799	5,733
	teacher	6,838	4,037	10,875
	Total	12,539	9,191	21,730
1997-98		Female	Male	
	ap	2,980	2,438	5,418
	prin	3,121	2,756	5,877
	teacher	6,820	4,097	10,917
	Total	12,921	9,291	22,212
1998-99		Female	Male	
	ap	3,183	2,537	5,720
	prin	3,258	2,756	6,014
	teacher	7,101	4,220	11,321
	Total	13,542	9,513	23,055
1999-		Female	Male	

00	ap	3,354	2,678	6,032
	prin	3,342	2,775	6,117
	teacher	7,136	4,128	11,264
	Total	13,832	9,581	23,413
2000-01		Female	Male	
	ap	3,604	2,722	6,326
	prin	3,473	2,780	6,253
	teacher	6,961	3,989	10,950
	Total	14,038	9,491	23,529
2001-02		Female	Male	
	ap	3,786	2,765	6,551
	prin	3,577	2,740	6,317
	teacher	7,162	4,126	11,288
	Total	14,525	9,631	24,156
2002-03		Female	Male	
	ap	4,007	2,841	6,848
	prin	3,697	2,650	6,347
	teacher	7,283	4,152	11,435
	Total	14,987	9,643	24,630
2003-04		Female	Male	
	ap	4,117	2,866	6,983
	prin	3,794	2,642	6,436
	teacher	7,694	4,305	11,999
	Total	15,605	9,813	25,418
2004-05		Female	Male	
	ap	4,383	2,922	7,305
	prin	3,892	2,691	6,583
	teacher	7,832	4,242	12,074
	Total	16,107	9,855	25,962
2005-06		Female	Male	
	ap	4,609	3,002	7,611
	prin	3,996	2,653	6,649
	teacher	7,876	4,132	12,008
	Total	16,481	9,787	26,268
2006-07		Female	Male	
	ap	4,934	3,080	8,014
	prin	4,104	2,687	6,791
	teacher	8,089	4,139	12,228
	Total	17,127	9,906	27,033
2007-08		Female	Male	
	ap	5,279	3,193	8,472
	prin	4,220	2,663	6,883
	teacher	8,257	4,162	12,419

	Total	17,756	10,018	27,774
2008-09		Female	Male	
	ap	5,496	3,211	8,707
	prin	4,382	2,641	7,023
	teacher	8,859	4,314	13,173
	Total	18,737	10,166	28,903
2009-10		Female	Male	
	ap	5,583	3,235	8,818
	prin	4,537	2,647	7,184
	teacher	9,633	4,536	14,169
	Total	19,753	10,418	30,171
2010-11		Female	Male	
	ap	5,607	3,299	8,906
	prin	4,604	2,660	7,264
	teacher	9,095	4,175	13,270
	Total	19,306	10,134	29,440

**Table A.6 - Frequencies of Role across Gender**

1991-92		Female	Male
	ap	48.74%	51.26%
	prin	41.33%	58.67%
	teacher	62.60%	37.40%
	Total	53.68%	46.32%
1992-93		Female	Male
	ap	50.25%	49.75%
	prin	43.98%	56.02%
	teacher	61.55%	38.45%
	Total	54.32%	45.68%
1993-94		Female	Male
	ap	50.56%	49.44%
	prin	46.20%	53.80%
	teacher	62.30%	37.70%
	Total	55.37%	44.63%
1994-95		Female	Male
	ap	51.81%	48.19%
	prin	48.19%	51.81%
	teacher	62.32%	37.68%
	Total	56.26%	43.74%
1995-96		Female	Male
	ap	52.97%	47.03%
	prin	49.60%	50.40%
	teacher	62.87%	37.13%
	Total	57.10%	42.90%
1996-97		Female	Male
	ap	54.02%	45.98%
	prin	51.18%	48.82%
	teacher	62.88%	37.12%
	Total	57.70%	42.30%
1997-98		Female	Male
	ap	55.00%	45.00%
	prin	53.11%	46.89%
	teacher	62.47%	37.53%
	Total	58.17%	41.83%
1998-99		Female	Male
	ap	55.65%	44.35%
	prin	54.17%	45.83%
	teacher	62.72%	37.28%
	Total	58.74%	41.26%
1999-		Female	Male

00	ap	55.60%	44.40%
	prin	54.63%	45.37%
	teacher	63.35%	36.65%
	Total	59.08%	40.92%
2000-01		Female	Male
	ap	56.97%	43.03%
	prin	55.54%	44.46%
	teacher	63.57%	36.43%
	Total	59.66%	40.34%
2001-02		Female	Male
	ap	57.79%	42.21%
	prin	56.62%	43.38%
	teacher	63.45%	36.55%
	Total	60.13%	39.87%
2002-03		Female	Male
	ap	58.51%	41.49%
	prin	58.25%	41.75%
	teacher	63.69%	36.31%
	Total	60.85%	39.15%
2003-04		Female	Male
	ap	58.96%	41.04%
	prin	58.95%	41.05%
	teacher	64.12%	35.88%
	Total	61.39%	38.61%
2004-05		Female	Male
	ap	60.00%	40.00%
	prin	59.12%	40.88%
	teacher	64.87%	35.13%
	Total	62.04%	37.96%
2005-06		Female	Male
	ap	60.56%	39.44%
	prin	60.10%	39.90%
	teacher	65.59%	34.41%
	Total	62.74%	37.26%
2006-07		Female	Male
	ap	61.57%	38.43%
	prin	60.43%	39.57%
	teacher	66.15%	33.85%
	Total	63.36%	36.64%
2007-08		Female	Male
	ap	62.31%	37.69%
	prin	61.31%	38.69%
	teacher	66.49%	33.51%

	Total	63.93%	36.07%
2008-09		Female	Male
	ap	63.12%	36.88%
	prin	62.39%	37.61%
	teacher	67.25%	32.75%
	Total	64.83%	35.17%
2009-10		Female	Male
	ap	63.31%	36.69%
	prin	63.15%	36.85%
	teacher	67.99%	32.01%
	Total	65.47%	34.53%
2010-11		Female	Male
	ap	62.96%	37.04%
	prin	63.38%	36.62%
	teacher	68.54%	31.46%
	Total	65.58%	34.42%

**Table A.7 - Frequencies of Role within Gender**

1991-92		Female	Male
	ap	19.84%	24.18%
	prin	21.34%	35.10%
	teacher	58.82%	40.72%
	Total	100.00%	100.00%
1992-93		Female	Male
	ap	20.38%	23.99%
	prin	21.88%	33.12%
	teacher	57.74%	42.88%
	Total	100.00%	100.00%
1993-94		Female	Male
	ap	20.14%	24.43%
	prin	22.50%	32.50%
	teacher	57.36%	43.07%
	Total	100.00%	100.00%
1994-95		Female	Male
	ap	20.57%	24.61%
	prin	22.51%	31.12%
	teacher	56.92%	44.27%
	Total	100.00%	100.00%
1995-96		Female	Male
	ap	21.34%	25.22%
	prin	22.86%	30.92%
	teacher	55.80%	43.86%
	Total	100.00%	100.00%
1996-97		Female	Male
	ap	22.07%	25.62%
	prin	23.40%	30.45%
	teacher	54.53%	43.92%
	Total	100.00%	100.00%
1997-98		Female	Male
	ap	23.06%	26.24%
	prin	24.15%	29.66%
	teacher	52.78%	44.10%
	Total	100.00%	100.00%
1998-99		Female	Male
	ap	23.50%	26.67%
	prin	24.06%	28.97%
	teacher	52.44%	44.36%
	Total	100.00%	100.00%
1999-		Female	Male

00	ap	24.25%	27.95%
	prin	24.16%	28.96%
	teacher	51.59%	43.09%
	Total	100.00%	100.00%
2000-01		Female	Male
	ap	25.67%	28.68%
	prin	24.74%	29.29%
	teacher	49.59%	42.03%
	Total	100.00%	100.00%
2001-02		Female	Male
	ap	26.07%	28.71%
	prin	24.63%	28.45%
	teacher	49.31%	42.84%
	Total	100.00%	100.00%
2002-03		Female	Male
	ap	26.74%	29.46%
	prin	24.67%	27.48%
	teacher	48.60%	43.06%
	Total	100.00%	100.00%
2003-04		Female	Male
	ap	26.38%	29.21%
	prin	24.31%	26.92%
	teacher	49.30%	43.87%
	Total	100.00%	100.00%
2004-05		Female	Male
	ap	27.21%	29.65%
	prin	24.16%	27.31%
	teacher	48.62%	43.04%
	Total	100.00%	100.00%
2005-06		Female	Male
	ap	27.97%	30.67%
	prin	24.25%	27.11%
	teacher	47.79%	42.22%
	Total	100.00%	100.00%
2006-07		Female	Male
	ap	28.81%	31.09%
	prin	23.96%	27.12%
	teacher	47.23%	41.78%
	Total	100.00%	100.00%
2007-08		Female	Male
	ap	29.73%	31.87%
	prin	23.77%	26.58%
	teacher	46.50%	41.55%



	Total	100.00%	100.00%
2008-09		Female	Male
	ap	29.33%	31.59%
	prin	23.39%	25.98%
	teacher	47.28%	42.44%
	Total	100.00%	100.00%
2009-10		Female	Male
	ap	28.26%	31.05%
	prin	22.97%	25.41%
	teacher	48.77%	43.54%
	Total	100.00%	100.00%
2010-11		Female	Male
	ap	29.04%	32.55%
	prin	23.85%	26.25%
	teacher	47.11%	41.20%
	Total	100.00%	100.00%

**Table A.8 - Multinomial Logistic Regression Results**

1991-92	pg. 127
1992-93	pg. 147
1993-94	pg. 165
1994-95	pg. 183
1995-96	pg. 204
1996-97	pg. 231
1997-98	pg. 259
1998-99	pg. 286
1999-00	pg. 313
2000-01	pg. 340
2001-02	pg. 367
2002-03	pg. 395
2003-04	pg. 422
2004-05	pg. 450
2005-06	pg. 477
2006-07	pg. 504
2007-08	pg. 530
2008-09	pg. 557
2009-10	pg. 584
2010-11	pg. 610

**1991-1992**

```
statewide:
.char ethnicity[omit] "White"
. mlogit admin Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale male
age tenure if validcert1992a==1, rrr
```

```
Iteration 0: log likelihood = -19846.197
Iteration 1: log likelihood = -19163.245
Iteration 2: log likelihood = -19157.563
Iteration 3: log likelihood = -19157.553
Iteration 4: log likelihood = -19157.553
```

Multinomial logistic regression	Number of obs	=	19256
	LR chi2(22)	=	1377.29
	Prob > chi2	=	0.0000
Log likelihood = -19157.553	Pseudo R2	=	0.0347

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	Asian	.436495	.2342884	-1.54	0.122	.1524395 1.249859
	Black	.8892755	.0699332	-1.49	0.136	.7622497 1.03747
	Latino	1.121119	.081466	1.57	0.116	.9722974 1.292719
	Other	.7036409	.277988	-0.89	0.374	.324389 1.526286
	AsianMale	2.848278	1.787461	1.67	0.095	.8325371 9.744538
	BlackMale	1.816439	.216406	5.01	0.000	1.438172 2.294198
	LatinoMale	1.423529	.1463641	3.43	0.001	1.163718 1.741346
	OtherMale	1.293968	.6524932	0.51	0.609	.4816139 3.476549
	male	1.583787	.0712147	10.23	0.000	1.450183 1.729701

	age		.9702733	.0025921	-11.30	0.000	.9652061	.9753672
	tenure		1.013886	.0026226	5.33	0.000	1.008758	1.019039
-----								
3								
	Asian		.7854809	.2962774	-0.64	0.522	.3750325	1.645138
	Black		.7374034	.0586339	-3.83	0.000	.6309903	.8617626
	Latino		1.261752	.0882057	3.33	0.001	1.100193	1.447036
	Other		.3251662	.172459	-2.12	0.034	.1149879	.9195147
	AsianMale		.6710508	.3248825	-0.82	0.410	.2598093	1.73323
	BlackMale		.8493148	.1053615	-1.32	0.188	.6659984	1.083089
	LatinoMale		.7089476	.0707435	-3.45	0.001	.5830095	.8620901
	OtherMale		1.173456	.7461802	0.25	0.801	.3374444	4.080666
	male		2.487052	.1016105	22.30	0.000	2.295664	2.694396
	age		1.008648	.0024744	3.51	0.000	1.00381	1.01351
	tenure		1.032329	.0023519	13.97	0.000	1.027729	1.036949
-----								

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=19256)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
	Asian	2.976	2	0.226
	Black	15.545	2	0.000
	Latino	11.345	2	0.003
	Other	6.146	2	0.046
	AsianMale	4.841	2	0.089
	BlackMale	35.064	2	0.000
	LatinoMale	37.140	2	0.000
	OtherMale	0.291	2	0.865
	male	520.489	2	0.000
	age	187.599	2	0.000
	tenure	198.947	2	0.000
-----				

\*\*\*\* Wald tests for independent variables (N=19256)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
	Asian	2.529	2	0.282
	Black	15.079	2	0.001
	Latino	11.488	2	0.003
	Other	4.856	2	0.088
	AsianMale	4.365	2	0.113
	BlackMale	35.038	2	0.000
	LatinoMale	37.101	2	0.000
	OtherMale	0.287	2	0.866
	male	507.613	2	0.000
	age	181.795	2	0.000
	tenure	195.543	2	0.000
-----				

\*\*\*\* Hausman tests of IIA assumption (N=19256)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
-----					
2		-4.853	12	---	---
3		-4.526	12	---	---
-----					

Note: If chi2<0, the estimated model does not

meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=19256)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-4643.560	-4635.675	15.770	12	0.202	for Ho
3	-4177.240	-4166.958	20.564	12	0.057	for Ho

\*\*\*\* Wald tests for combining alternatives (N=19256)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2- 3	525.973	11	0.000
2- 1	426.013	11	0.000
3- 1	937.167	11	0.000

\*\*\*\* LR tests for combining alternatives (N=19256)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2- 3	553.936	11	0.000
2- 1	435.608	11	0.000
3- 1	995.647	11	0.000

by region

. bysort region: mlogit admin Asian Black Latino Other AsianMale BlackMale LatinoMale  
OtherMale male age tenure if validcert1992a==1, rrr

-----> region = 1

Iteration 0: log likelihood = -1272.2522  
Iteration 1: log likelihood = -1197.3667  
Iteration 2: log likelihood = -1196.279  
Iteration 3: log likelihood = -1196.0796  
Iteration 4: log likelihood = -1196.0357  
Iteration 5: log likelihood = -1196.025  
Iteration 6: log likelihood = -1196.0228  
Iteration 7: log likelihood = -1196.0223  
Iteration 8: log likelihood = -1196.0222  
Iteration 9: log likelihood = -1196.0222

Multinomial logistic regression                      Number of obs    =       1204  
   LR chi2(22)       =       152.46  
   Prob > chi2       =       0.0000  
Log likelihood = -1196.0222                          Pseudo R2        =       0.0599

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
Asian	2.30e-06	.0035335	-0.01	0.993	0 .
Black	1.49e+07	1.94e+10	0.01	0.990	0 .

	Latino		1.160508	.3002624	0.58	0.565	.6988982	1.927004
	Other		1.13e-06	.0017356	-0.01	0.993	0	.
	AsianMale		.420467	913.1905	-0.00	1.000	0	.
	BlackMale		5.38e-14	8.50e-11	-0.02	0.985	0	.
	LatinoMale		1.293741	.4615135	0.72	0.470	.6429836	2.603123
	OtherMale		3036197	4.66e+09	0.01	0.992	0	.
	male		1.810739	.5744831	1.87	0.061	.9722991	3.372189
	age		.9463085	.0105224	-4.96	0.000	.925908	.9671585
	tenure		1.0465	.0109257	4.35	0.000	1.025303	1.068134
-----								
3								
	Asian		6.55e-07	.0010187	-0.01	0.993	0	.
	Black		2.688707	5885.819	0.00	1.000	0	.
	Latino		1.033817	.2500918	0.14	0.891	.6434699	1.660958
	Other		8.75e-07	.0013614	-0.01	0.993	0	.
	AsianMale		2.003039	4405.541	0.00	1.000	0	.
	BlackMale		4.86e-07	.0011451	-0.01	0.995	0	.
	LatinoMale		1.830772	.6539185	1.69	0.090	.9090764	3.686956
	OtherMale		4054772	6.31e+09	0.01	0.992	0	.
	male		1.02766	.3242769	0.09	0.931	.5536735	1.907414
	age		.9964188	.0109201	-0.33	0.743	.975244	1.018053
	tenure		1.07272	.0107132	7.03	0.000	1.051927	1.093925
-----								

-----> region = 2

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -807.17651  
Iteration 1: log likelihood = -776.30991  
Iteration 2: log likelihood = -775.4422  
Iteration 3: log likelihood = -774.73356  
Iteration 4: log likelihood = -774.69385  
Iteration 5: log likelihood = -774.68682  
Iteration 6: log likelihood = -774.68528  
Iteration 7: log likelihood = -774.68489  
Iteration 8: log likelihood = -774.68482  
Iteration 9: log likelihood = -774.6848

Multinomial logistic regression	Number of obs	=	859
	LR chi2(18)	=	64.98
	Prob > chi2	=	0.0000
Log likelihood = -774.6848	Pseudo R2	=	0.0403

-----							
	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----							
1			(base outcome)				
-----							
2							
	Asian		(omitted)				
	Black		.802959	.6344426	-0.28	0.781	.1706604 3.77793
	Latino		.824396	.2296929	-0.69	0.488	.4775012 1.423303
	Other		9.386849	11.74794	1.79	0.074	.8076337 109.1001
	AsianMale		(omitted)				
	BlackMale		3.667959	3.960567	1.20	0.229	.4418951 30.44597
	LatinoMale		1.438581	.5778246	0.91	0.365	.6546935 3.161043
	OtherMale		6.70e-08	.0000866	-0.01	0.990	0 .
	male		1.490137	.4068584	1.46	0.144	.8726075 2.544681
	age		.997017	.01448	-0.21	0.837	.9690368 1.025805
	tenure		1.012715	.0142249	0.90	0.368	.985215 1.040983
-----							
3							
	Asian		(omitted)				
	Black		1.181563	.805765	0.24	0.807	.3104425 4.497099
	Latino		1.22284	.3143387	0.78	0.434	.7388614 2.023841

Other		8.86e-07	.0025963	-0.00	0.996	0	.
AsianMale		(omitted)					
BlackMale		.9475953	.9844905	-0.05	0.959	.1236742	7.2605
LatinoMale		.8521235	.2982462	-0.46	0.648	.4291202	1.6921
OtherMale		.6492682	2024.537	-0.00	1.000	0	.
male		2.726019	.6603589	4.14	0.000	1.695631	4.382545
age		1.026795	.0129496	2.10	0.036	1.001725	1.052492
tenure		1.019138	.0120045	1.61	0.108	.9958789	1.04294

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -332.76844  
Iteration 1: log likelihood = -313.27778  
Iteration 2: log likelihood = -312.90432  
Iteration 3: log likelihood = -312.87462  
Iteration 4: log likelihood = -312.8688  
Iteration 5: log likelihood = -312.86757  
Iteration 6: log likelihood = -312.8673  
Iteration 7: log likelihood = -312.86723  
Iteration 8: log likelihood = -312.86721

Multinomial logistic regression		Number of obs	=	324
		LR chi2(16)	=	39.80
		Prob > chi2	=	0.0008
Log likelihood = -312.86721		Pseudo R2	=	0.0598

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
Asian		(omitted)				
Black		2.086675	1.920844	0.80	0.424	.3434786 12.67681
Latino		4.061465	3.501274	1.63	0.104	.7496974 22.00288
Other		.6749088	975.6196	-0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		2.192741	3.796546	0.45	0.650	.0736553 65.2786
LatinoMale		.5226977	.6224857	-0.54	0.586	.0506461 5.394544
OtherMale		(omitted)				
male		1.371477	.4928605	0.88	0.379	.6781049 2.773834
age		1.007603	.0273987	0.28	0.781	.955308 1.06276
tenure		.9485571	.0215394	-2.33	0.020	.9072664 .991727
3						
Asian		(omitted)				
Black		1.341573	1.206458	0.33	0.744	.2302187 7.817863
Latino		1.576494	1.480531	0.48	0.628	.2502096 9.933
Other		489194.8	3.71e+08	0.02	0.986	0 .
AsianMale		(omitted)				
BlackMale		1.078057	1.658747	0.05	0.961	.0528379 21.99573
LatinoMale		1.295427	1.464282	0.23	0.819	.1413382 11.87316
OtherMale		(omitted)				
male		3.152539	.8888116	4.07	0.000	1.814165 5.478279
age		1.030995	.0225647	1.39	0.163	.9877044 1.076183
tenure		.9876306	.0161855	-0.76	0.448	.9564116 1.019869

-----> region = 4

--Break--

```

r(1);

. bysort region: mlogit admin Asian Black Latino Other AsianMale BlackMale LatinoMale
OtherMale male age tenure if validcert1992a==1, rrr iter(20)

```

```

-----> region = 1

```

```

Iteration 0: log likelihood = -1272.2522
Iteration 1: log likelihood = -1197.3667
Iteration 2: log likelihood = -1196.279
Iteration 3: log likelihood = -1196.0796
Iteration 4: log likelihood = -1196.0357
Iteration 5: log likelihood = -1196.025
Iteration 6: log likelihood = -1196.0228
Iteration 7: log likelihood = -1196.0223
Iteration 8: log likelihood = -1196.0222
Iteration 9: log likelihood = -1196.0222

```

```

Multinomial logistic regression      Number of obs   =      1204
LR chi2(22)                        =      152.46
Prob > chi2                        =      0.0000
Pseudo R2                          =      0.0599
Log likelihood = -1196.0222

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	Asian	2.30e-06	.0035335	-0.01	0.993	0	.
	Black	1.49e+07	1.94e+10	0.01	0.990	0	.
	Latino	1.160508	.3002624	0.58	0.565	.6988982	1.927004
	Other	1.13e-06	.0017356	-0.01	0.993	0	.
	AsianMale	.420467	913.1905	-0.00	1.000	0	.
	BlackMale	5.38e-14	8.50e-11	-0.02	0.985	0	.
	LatinoMale	1.293741	.4615135	0.72	0.470	.6429836	2.603123
	OtherMale	3036197	4.66e+09	0.01	0.992	0	.
	male	1.810739	.5744831	1.87	0.061	.9722991	3.372189
	age	.9463085	.0105224	-4.96	0.000	.925908	.9671585
	tenure	1.0465	.0109257	4.35	0.000	1.025303	1.068134
3							
	Asian	6.55e-07	.0010187	-0.01	0.993	0	.
	Black	2.688707	5885.819	0.00	1.000	0	.
	Latino	1.033817	.2500918	0.14	0.891	.6434699	1.660958
	Other	8.75e-07	.0013614	-0.01	0.993	0	.
	AsianMale	2.003039	4405.541	0.00	1.000	0	.
	BlackMale	4.86e-07	.0011451	-0.01	0.995	0	.
	LatinoMale	1.830772	.6539185	1.69	0.090	.9090764	3.686956
	OtherMale	4054772	6.31e+09	0.01	0.992	0	.
	male	1.02766	.3242769	0.09	0.931	.5536735	1.907414
	age	.9964188	.0109201	-0.33	0.743	.975244	1.018053
	tenure	1.07272	.0107132	7.03	0.000	1.051927	1.093925

```

-----> region = 2

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -807.17651
Iteration 1: log likelihood = -776.30991
Iteration 2: log likelihood = -775.4422
Iteration 3: log likelihood = -774.73356
Iteration 4: log likelihood = -774.69385
Iteration 5: log likelihood = -774.68682

```

```
Iteration 6: log likelihood = -774.68528
Iteration 7: log likelihood = -774.68489
Iteration 8: log likelihood = -774.68482
Iteration 9: log likelihood = -774.6848
```

```
Multinomial logistic regression      Number of obs   =      859
                                     LR chi2(18)        =      64.98
                                     Prob > chi2         =      0.0000
Log likelihood = -774.6848           Pseudo R2        =      0.0403
```

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)					
2							
	Asian	(omitted)					
	Black		.802959	.6344426	-0.28	0.781	.1706604 3.77793
	Latino		.824396	.2296929	-0.69	0.488	.4775012 1.423303
	Other		9.386849	11.74794	1.79	0.074	.8076337 109.1001
	AsianMale	(omitted)					
	BlackMale		3.667959	3.960567	1.20	0.229	.4418951 30.44597
	LatinoMale		1.438581	.5778246	0.91	0.365	.6546935 3.161043
	OtherMale		6.70e-08	.0000866	-0.01	0.990	0 .
	male		1.490137	.4068584	1.46	0.144	.8726075 2.544681
	age		.997017	.01448	-0.21	0.837	.9690368 1.025805
	tenure		1.012715	.0142249	0.90	0.368	.985215 1.040983
3							
	Asian	(omitted)					
	Black		1.181563	.805765	0.24	0.807	.3104425 4.497099
	Latino		1.22284	.3143387	0.78	0.434	.7388614 2.023841
	Other		8.86e-07	.0025963	-0.00	0.996	0 .
	AsianMale	(omitted)					
	BlackMale		.9475953	.9844905	-0.05	0.959	.1236742 7.2605
	LatinoMale		.8521235	.2982462	-0.46	0.648	.4291202 1.6921
	OtherMale		.6492682	2024.537	-0.00	1.000	0 .
	male		2.726019	.6603589	4.14	0.000	1.695631 4.382545
	age		1.026795	.0129496	2.10	0.036	1.001725 1.052492
	tenure		1.019138	.0120045	1.61	0.108	.9958789 1.04294

```
-----> region = 3
```

```
note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -332.76844
Iteration 1: log likelihood = -313.27778
Iteration 2: log likelihood = -312.90432
Iteration 3: log likelihood = -312.87462
Iteration 4: log likelihood = -312.8688
Iteration 5: log likelihood = -312.86757
Iteration 6: log likelihood = -312.8673
Iteration 7: log likelihood = -312.86723
Iteration 8: log likelihood = -312.86721
```

```
Multinomial logistic regression      Number of obs   =      324
                                     LR chi2(16)        =      39.80
                                     Prob > chi2         =      0.0008
Log likelihood = -312.86721           Pseudo R2        =      0.0598
```

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)					



-----+-----							
2							
	Asian	(omitted)					
	Black	2.086675	1.920844	0.80	0.424	.3434786	12.67681
	Latino	4.061465	3.501274	1.63	0.104	.7496974	22.00288
	Other	.6749088	975.6196	-0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	2.192741	3.796546	0.45	0.650	.0736553	65.2786
	LatinoMale	.5226977	.6224857	-0.54	0.586	.0506461	5.394544
	OtherMale	(omitted)					
	male	1.371477	.4928605	0.88	0.379	.6781049	2.773834
	age	1.007603	.0273987	0.28	0.781	.955308	1.06276
	tenure	.9485571	.0215394	-2.33	0.020	.9072664	.991727
-----+-----							

3							
	Asian	(omitted)					
	Black	1.341573	1.206458	0.33	0.744	.2302187	7.817863
	Latino	1.576494	1.480531	0.48	0.628	.2502096	9.933
	Other	489194.8	3.71e+08	0.02	0.986	0	.
	AsianMale	(omitted)					
	BlackMale	1.078057	1.658747	0.05	0.961	.0528379	21.99573
	LatinoMale	1.295427	1.464282	0.23	0.819	.1413382	11.87316
	OtherMale	(omitted)					
	male	3.152539	.8888116	4.07	0.000	1.814165	5.478279
	age	1.030995	.0225647	1.39	0.163	.9877044	1.076183
	tenure	.9876306	.0161855	-0.76	0.448	.9564116	1.019869
-----+-----							

-----> region = 4

Iteration 0: log likelihood = -3949.4578  
Iteration 1: log likelihood = -3734.3365  
Iteration 2: log likelihood = -3728.6781  
Iteration 3: log likelihood = -3728.4779  
Iteration 4: log likelihood = -3728.4301  
Iteration 5: log likelihood = -3728.4204  
Iteration 6: log likelihood = -3728.4183  
Iteration 7: log likelihood = -3728.4178  
Iteration 8: log likelihood = -3728.4177  
Iteration 9: log likelihood = -3728.4177

Multinomial logistic regression	Number of obs	=	3879
	LR chi2(22)	=	442.08
	Prob > chi2	=	0.0000
Log likelihood = -3728.4177	Pseudo R2	=	0.0560

-----+-----							
	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
-----+-----							
2							
	Asian	.6301313	.518843	-0.56	0.575	.1254762	3.164468
	Black	.4585693	.0572969	-6.24	0.000	.3589632	.5858145
	Latino	1.252405	.2526741	1.12	0.265	.8433623	1.859841
	Other	.448598	.5048009	-0.71	0.476	.0494325	4.071009
	AsianMale	1.76e-06	.0008751	-0.03	0.979	0	.
	BlackMale	2.333515	.4650226	4.25	0.000	1.579004	3.448561
	LatinoMale	1.571271	.5143033	1.38	0.167	.8272532	2.984445
	OtherMale	.6031	.9387153	-0.32	0.745	.0285436	12.74297
	male	1.134321	.1052808	1.36	0.174	.9456546	1.360628
	age	.9623738	.0053713	-6.87	0.000	.9519037	.9729591
	tenure	1.024146	.0058763	4.16	0.000	1.012693	1.035728
-----+-----							
3							
	Asian	.6414229	.6971422	-0.41	0.683	.076208	5.398689

Black		.3122562	.0472305	-7.70	0.000	.2321466	.4200103
Latino		2.094742	.4417456	3.51	0.000	1.385564	3.166901
Other		.9977482	1.136988	-0.00	0.998	.1069131	9.311318
AsianMale		.3935521	.6083872	-0.60	0.546	.0190169	8.144492
BlackMale		2.070428	.4835849	3.12	0.002	1.309927	3.272452
LatinoMale		.8666393	.3136247	-0.40	0.692	.4263834	1.761475
OtherMale		.5939355	.840025	-0.37	0.713	.037141	9.497849
male		1.323702	.1386444	2.68	0.007	1.078042	1.625343
age		.9961917	.0064954	-0.59	0.558	.9835419	1.009004
tenure		1.087037	.006909	13.13	0.000	1.07358	1.100663

-----> region = 5

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -568.03994  
Iteration 1: log likelihood = -507.29957  
Iteration 2: log likelihood = -505.12232  
Iteration 3: log likelihood = -504.85699  
Iteration 4: log likelihood = -504.82732  
Iteration 5: log likelihood = -504.82303  
Iteration 6: log likelihood = -504.82227  
Iteration 7: log likelihood = -504.8221  
Iteration 8: log likelihood = -504.82206  
Iteration 9: log likelihood = -504.82205

Multinomial logistic regression	Number of obs	=	619
	LR chi2(18)	=	126.44
	Prob > chi2	=	0.0000
Log likelihood = -504.82205	Pseudo R2	=	0.1113

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
Asian		(omitted)				
Black		2.178729	.9202306	1.84	0.065	.9520962 4.985693
Latino		4.362391	5.288306	1.22	0.224	.4053741 46.94541
Other		6.82e-07	.002535	-0.00	0.997	0 .
AsianMale		(omitted)				
BlackMale		.5086219	.2909236	-1.18	0.237	.1657759 1.560518
LatinoMale		2.43e-08	.0000516	-0.01	0.993	0 .
OtherMale		.1251797	570.1363	-0.00	1.000	0 .
male		6.838185	2.084887	6.31	0.000	3.762001 12.42976
age		.9444582	.0169087	-3.19	0.001	.9118925 .9781869
tenure		1.042065	.0172314	2.49	0.013	1.008834 1.076391
3						
Asian		(omitted)				
Black		1.456785	.5897255	0.93	0.353	.6588977 3.220868
Latino		1.74e-06	.0025945	-0.01	0.993	0 .
Other		4.33e-07	.0014017	-0.00	0.996	0 .
AsianMale		(omitted)				
BlackMale		.4868219	.2701436	-1.30	0.195	.1640694 1.444483
LatinoMale		.054924	129.9643	-0.00	0.999	0 .
OtherMale		.3013888	1188.032	-0.00	1.000	0 .
male		5.719688	1.448552	6.89	0.000	3.481765 9.396048
age		1.010088	.014914	0.68	0.497	.9812763 1.039746
tenure		1.034915	.0140007	2.54	0.011	1.007835 1.062723

-----> region = 6

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -659.93438
Iteration 1: log likelihood = -630.59529
Iteration 2: log likelihood = -629.64267
Iteration 3: log likelihood = -629.45648
Iteration 4: log likelihood = -629.41521
Iteration 5: log likelihood = -629.40504
Iteration 6: log likelihood = -629.403
Iteration 7: log likelihood = -629.40257
Iteration 8: log likelihood = -629.40247
Iteration 9: log likelihood = -629.40244
```

```
Multinomial logistic regression      Number of obs   =      622
                                     LR chi2(20)        =      61.06
                                     Prob > chi2         =      0.0000
Log likelihood = -629.40244          Pseudo R2        =      0.0463
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	6.62e-07	.0011051	-0.01	0.993	0	.
	Black	.431744	.280928	-1.29	0.197	.1206049	1.545566
	Latino	1.876913	2.672978	0.44	0.658	.1151377	30.59642
	Other	1.64e-06	.002743	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	9.887005	9.487537	2.39	0.017	1.507491	64.84476
	LatinoMale	7.40e-07	.0005575	-0.02	0.985	0	.
	OtherMale	1.101393	2115.555	0.00	1.000	0	.
	male	1.622099	.3577269	2.19	0.028	1.052832	2.49917
	age	.9661609	.014195	-2.34	0.019	.9387359	.9943871
	tenure	1.002821	.0139574	0.20	0.840	.9758343	1.030553
3							
	Asian	1.19e-06	.0017371	-0.01	0.993	0	.
	Black	.3255155	.2497204	-1.46	0.143	.0723718	1.464111
	Latino	2.98e-06	.0027442	-0.01	0.989	0	.
	Other	1.01e-06	.0014666	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	5.794391	6.052515	1.68	0.093	.747976	44.88776
	LatinoMale	136011.7	1.25e+08	0.01	0.990	0	.
	OtherMale	809885.7	1.18e+09	0.01	0.993	0	.
	male	2.885804	.5838956	5.24	0.000	1.941065	4.290359
	age	1.006556	.0130778	0.50	0.615	.9812479	1.032517
	tenure	1.000131	.0121843	0.01	0.991	.9765335	1.0243

-----> region = 7

```
Iteration 0: log likelihood = -965.44535
Iteration 1: log likelihood = -900.16614
Iteration 2: log likelihood = -895.07265
Iteration 3: log likelihood = -894.85025
Iteration 4: log likelihood = -894.79708
Iteration 5: log likelihood = -894.7848
Iteration 6: log likelihood = -894.78174
Iteration 7: log likelihood = -894.78115
Iteration 8: log likelihood = -894.78105
Iteration 9: log likelihood = -894.78103
```

```
Multinomial logistic regression      Number of obs   =     1009
                                     LR chi2(22)        =     141.33
                                     Prob > chi2         =      0.0000
```

Log likelihood = -894.78103

Pseudo R2 = 0.0732

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
Asian	5.47e+08	4.27e+12	0.00	0.998	0	.
Black	.9766125	.4976903	-0.05	0.963	.3597017	2.651564
Latino	3.47e-06	.0026285	-0.02	0.987	0	.
Other	2.71e-06	.0056986	-0.01	0.995	0	.
AsianMale	.008687	68.68999	-0.00	1.000	0	.
BlackMale	2.315818	1.398344	1.39	0.164	.709136	7.562745
LatinoMale	.2632608	337.0248	-0.00	0.999	0	.
OtherMale	8.21e+13	6.64e+17	0.00	0.997	0	.
male	4.165261	.986823	6.02	0.000	2.618049	6.626843
age	.9732385	.0140495	-1.88	0.060	.946088	1.001168
tenure	1.02511	.0142566	1.78	0.075	.9975448	1.053437
3						
Asian	.875689	11478.68	-0.00	1.000	0	.
Black	.9089616	.2951558	-0.29	0.769	.4810004	1.717693
Latino	.7590419	.839932	-0.25	0.803	.0867666	6.640162
Other	1.82e-06	.0025227	-0.01	0.992	0	.
AsianMale	1185630	1.56e+10	0.00	0.999	0	.
BlackMale	.6245453	.2971986	-0.99	0.323	.245758	1.587158
LatinoMale	.6496004	1.077373	-0.26	0.795	.0251718	16.76402
OtherMale	308142.7	4.06e+09	0.00	0.999	0	.
male	3.337378	.5366886	7.49	0.000	2.435131	4.573918
age	1.015709	.0106295	1.49	0.136	.9950874	1.036757
tenure	1.021326	.0100613	2.14	0.032	1.001795	1.041237

-----> region = 8

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -288.99157  
Iteration 1: log likelihood = -260.23858  
Iteration 2: log likelihood = -255.59635  
Iteration 3: log likelihood = -254.98773  
Iteration 4: log likelihood = -254.9271  
Iteration 5: log likelihood = -254.91441  
Iteration 6: log likelihood = -254.91153  
Iteration 7: log likelihood = -254.91082  
Iteration 8: log likelihood = -254.91068  
Iteration 9: log likelihood = -254.91066

Multinomial logistic regression      Number of obs      =      299  
LR chi2(16)      =      68.16  
Prob > chi2      =      0.0000  
Log likelihood = -254.91066      Pseudo R2      =      0.1179

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	(omitted)					
	Black	25.13485	38.65189	2.10	0.036	1.234	511.962
	Latino	.0000213	.0278722	-0.01	0.993	0	.
	Other	.0000232	.0304575	-0.01	0.994	0	.
	AsianMale	(omitted)					

BlackMale		.3317377	.5794673	-0.63	0.528	.010813	10.17754
LatinoMale		.0288965	65.6215	-0.00	0.999	0	.
OtherMale		(omitted)					
male		31.9317	33.01269	3.35	0.001	4.20925	242.2364
age		1.01072	.0278866	0.39	0.699	.9575148	1.066882
tenure		.996975	.0272738	-0.11	0.912	.9449271	1.05189
-----							
3							
Asian		(omitted)					
Black		3.261373	2.620701	1.47	0.141	.6751665	15.75397
Latino		2.83e-06	.0020361	-0.02	0.986	0	.
Other		2.73e-06	.0019678	-0.02	0.986	0	.
AsianMale		(omitted)					
BlackMale		.9146301	1.043639	-0.08	0.938	.0977196	8.5607
LatinoMale		.2681093	334.4935	-0.00	0.999	0	.
OtherMale		(omitted)					
male		3.609151	1.026619	4.51	0.000	2.06672	6.302726
age		1.009782	.0182565	0.54	0.590	.9746267	1.046206
tenure		1.02273	.0183826	1.25	0.211	.9873279	1.059401
-----							

-----> region = 9

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -212.42355  
Iteration 1: log likelihood = -196.30347  
Iteration 2: log likelihood = -195.41904  
Iteration 3: log likelihood = -195.29447  
Iteration 4: log likelihood = -195.26733  
Iteration 5: log likelihood = -195.26164  
Iteration 6: log likelihood = -195.26068  
Iteration 7: log likelihood = -195.26045  
Iteration 8: log likelihood = -195.2604  
Iteration 9: log likelihood = -195.26039

Multinomial logistic regression	Number of obs	=	209
	LR chi2(14)	=	34.33
	Prob > chi2	=	0.0018
Log likelihood = -195.26039	Pseudo R2	=	0.0808

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	Asian	(omitted)					
	Black	.4710064	.6027912	-0.59	0.556	.0383411	5.786146
	Latino	2.52e-07	.0006504	-0.01	0.995	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	9.049876	14.39702	1.38	0.166	.4004116	204.5402
	LatinoMale	6672542	9.98e+10	0.00	0.999	0	.
	OtherMale	(omitted)					
	male	.3262062	.1114153	-3.28	0.001	.1670188	.6371169
	age	.9477301	.0243689	-2.09	0.037	.9011515	.9967163
	tenure	.9935108	.0218977	-0.30	0.768	.951506	1.03737
2							
	Asian	(omitted)					
	Black	5.293435	5.498646	1.60	0.109	.6910946	40.54504
	Latino	2.18e-07	.0008752	-0.00	0.997	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	1.54e-07	.0002206	-0.01	0.991	0	.

LatinoMale		1.58e+15	1.69e+19	0.00	0.997	0	.
OtherMale		(omitted)					
male		1.036956	.5171665	0.07	0.942	.3901543	2.756034
age		.9675503	.0333508	-0.96	0.339	.9043431	1.035175
tenure		1.01761	.0294401	0.60	0.546	.9615138	1.076979

-----+-----  
3 | (base outcome)  
-----+-----

-----> region = 10

Iteration 0: log likelihood = -2585.0173  
Iteration 1: log likelihood = -2488.5036  
Iteration 2: log likelihood = -2487.0179  
Iteration 3: log likelihood = -2486.7626  
Iteration 4: log likelihood = -2486.7232  
Iteration 5: log likelihood = -2486.7137  
Iteration 6: log likelihood = -2486.7117  
Iteration 7: log likelihood = -2486.7113  
Iteration 8: log likelihood = -2486.7112  
Iteration 9: log likelihood = -2486.7112

Multinomial logistic regression	Number of obs	=	2565
	LR chi2(22)	=	196.61
	Prob > chi2	=	0.0000
Log likelihood = -2486.7112	Pseudo R2	=	0.0380

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	Asian	.3966634	.4240919	-0.86	0.387	.0487935 3.224649
	Black	1.070918	.2011968	0.36	0.715	.7410345 1.547655
	Latino	1.786598	.5463875	1.90	0.058	.9810867 3.253465
	Other	.8379344	.6618389	-0.22	0.823	.1781936 3.940287
	AsianMale	2.465864	2.829197	0.79	0.431	.2602239 23.36636
	BlackMale	1.451722	.4144059	1.31	0.192	.8296615 2.540188
	LatinoMale	.937671	.4289448	-0.14	0.888	.3825268 2.298472
	OtherMale	1.21791	1.299523	0.18	0.853	.1504439 9.859523
	male	1.9275	.2384535	5.30	0.000	1.512485 2.456392
	age	.9743153	.0070064	-3.62	0.000	.9606794 .9881448
	tenure	1.028737	.0076074	3.83	0.000	1.013934 1.043756

3						
	Asian	1.208871	.6540175	0.35	0.726	.4186653 3.490542
	Black	.9563249	.170528	-0.25	0.802	.6742527 1.356402
	Latino	2.23175	.6036341	2.97	0.003	1.313458 3.792057
	Other	1.56e-06	.0006738	-0.03	0.975	0 .
	AsianMale	.5061806	.3292229	-1.05	0.295	.1414747 1.811057
	BlackMale	.7741645	.2247509	-0.88	0.378	.4382454 1.367569
	LatinoMale	.6396766	.269119	-1.06	0.288	.2804469 1.459051
	OtherMale	340766	1.48e+08	0.03	0.977	0 .
	male	2.251201	.2494613	7.32	0.000	1.811718 2.797294
	age	1.000215	.0065851	0.03	0.974	.9873916 1.013205
	tenure	1.048977	.0068902	7.28	0.000	1.035559 1.062569

-----> region = 11

Iteration 0: log likelihood = -1812.7252  
Iteration 1: log likelihood = -1726.5652  
Iteration 2: log likelihood = -1724.5621  
Iteration 3: log likelihood = -1724.1848

```

Iteration 4: log likelihood = -1724.1229
Iteration 5: log likelihood = -1724.1165
Iteration 6: log likelihood = -1724.1151
Iteration 7: log likelihood = -1724.1148
Iteration 8: log likelihood = -1724.1148
Iteration 9: log likelihood = -1724.1148

```

```

Multinomial logistic regression      Number of obs   =      1710
                                     LR chi2(22)        =      177.22
                                     Prob > chi2         =      0.0000
Log likelihood = -1724.1148          Pseudo R2        =      0.0489

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	2.42e-07	.000314	-0.01	0.991	0	.
	Black	1.613103	.4453935	1.73	0.083	.9389381	2.771323
	Latino	1.53276	.5575417	1.17	0.240	.7513623	3.126791
	Other	1.278424	1.146895	0.27	0.784	.2203173	7.418251
	AsianMale	8.07e+13	2.79e+17	0.01	0.993	0	.
	BlackMale	1.324007	.5791495	0.64	0.521	.5617667	3.120501
	LatinoMale	1.637794	.9444074	0.86	0.392	.5289661	5.070968
	OtherMale	1.130657	1.530747	0.09	0.928	.0796024	16.05963
	male	1.250359	.1691439	1.65	0.099	.9591519	1.629979
	age	.9780461	.0080956	-2.68	0.007	.9623071	.9940426
	tenure	.9572532	.0078125	-5.35	0.000	.9420627	.9726886
3							
	Asian	5.56e-07	.000709	-0.01	0.991	0	.
	Black	1.784196	.5114599	2.02	0.043	1.01727	3.129314
	Latino	2.159433	.7998906	2.08	0.038	1.044817	4.463127
	Other	1.03e-06	.0008759	-0.02	0.987	0	.
	AsianMale	811560.3	4.40e+09	0.00	0.998	0	.
	BlackMale	.2990303	.1519382	-2.38	0.018	.1104631	.8094933
	LatinoMale	.5804187	.3498756	-0.90	0.367	.1780895	1.891666
	OtherMale	.3401155	470.815	-0.00	0.999	0	.
	male	2.371733	.3028838	6.76	0.000	1.846556	3.046275
	age	1.034387	.0083643	4.18	0.000	1.018122	1.050911
	tenure	.9757106	.0070712	-3.39	0.001	.9619492	.9896689

```

-----> region = 12

```

```

note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -670.5475
Iteration 1: log likelihood = -638.59926
Iteration 2: log likelihood = -636.29607
Iteration 3: log likelihood = -635.71901
Iteration 4: log likelihood = -635.63032
Iteration 5: log likelihood = -635.61215
Iteration 6: log likelihood = -635.60833
Iteration 7: log likelihood = -635.60745
Iteration 8: log likelihood = -635.60724
Iteration 9: log likelihood = -635.6072
Iteration 10: log likelihood = -635.60719

```

```

Multinomial logistic regression      Number of obs   =      664
                                     LR chi2(18)        =      69.88
                                     Prob > chi2         =      0.0000
Log likelihood = -635.60719          Pseudo R2        =      0.0521

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	2.283362	3.267251	0.58	0.564	.1382277	37.71851
	Black	.8927793	.480977	-0.21	0.833	.3105752	2.566383
	Latino	8.38e-07	.0007712	-0.02	0.988	0	.
	Other	2.40e+08	2.16e+12	0.00	0.998	0	.
	AsianMale	(omitted)					
	BlackMale	2.428845	1.762413	1.22	0.221	.5858047	10.0704
	LatinoMale	4508803	4.15e+09	0.02	0.987	0	.
	OtherMale	(omitted)					
	male	1.566809	.3938183	1.79	0.074	.9573395	2.564283
	age	.9855959	.0158391	-0.90	0.367	.9550356	1.017134
	tenure	1.032043	.0165517	1.97	0.049	1.000107	1.064999
3							
	Asian	1.48e-07	.0003066	-0.01	0.994	0	.
	Black	1.140473	.4747137	0.32	0.752	.5044039	2.578645
	Latino	1.768491	1.183388	0.85	0.394	.4764533	6.564252
	Other	1.000476	13899.91	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.5918031	.3737674	-0.83	0.406	.171625	2.040676
	LatinoMale	8.40e-08	.0001207	-0.01	0.991	0	.
	OtherMale	(omitted)					
	male	3.028613	.5946807	5.64	0.000	2.061136	4.450212
	age	1.009688	.0121445	0.80	0.423	.986164	1.033774
	tenure	1.034573	.0124645	2.82	0.005	1.010429	1.059294

-----> region = 13

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1053.1728  
Iteration 1: log likelihood = -1021.8579  
Iteration 2: log likelihood = -1021.1967  
Iteration 3: log likelihood = -1021.0507  
Iteration 4: log likelihood = -1021.0273  
Iteration 5: log likelihood = -1021.0219  
Iteration 6: log likelihood = -1021.0206  
Iteration 7: log likelihood = -1021.0203  
Iteration 8: log likelihood = -1021.0203

Multinomial logistic regression	Number of obs	=	978
	LR chi2(18)	=	64.30
	Prob > chi2	=	0.0000
Log likelihood = -1021.0203	Pseudo R2	=	0.0305

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	(omitted)					
	Black	1.725125	.6531507	1.44	0.150	.821385	3.623219
	Latino	1.773629	.5827342	1.74	0.081	.9315292	3.376985
	Other	5.03e-06	.0031805	-0.02	0.985	0	.
	AsianMale	(omitted)					
	BlackMale	2.054111	1.365073	1.08	0.279	.5584089	7.556062
	LatinoMale	1.068456	.5451071	0.13	0.897	.3930872	2.904186
	OtherMale	166097.8	1.05e+08	0.02	0.985	0	.
	male	1.766713	.3328197	3.02	0.003	1.221275	2.55575
	age	.9641189	.0116896	-3.01	0.003	.9414777	.9873046



	tenure		1.031294	.0132554	2.40	0.017	1.005639	1.057604
3	Asian		(omitted)					
	Black		1.382379	.5176103	0.86	0.387	.6636026	2.879694
	Latino		1.82622	.5630248	1.95	0.051	.9979921	3.341789
	Other		1.138553	1.410801	0.10	0.917	.1003726	12.91492
	AsianMale		(omitted)					
	BlackMale		1.339182	.9045857	0.43	0.665	.3563472	5.032758
	LatinoMale		.6493582	.3219938	-0.87	0.384	.2456991	1.716189
	OtherMale		1.24e-06	.0006952	-0.02	0.981	0	.
	male		2.080143	.352708	4.32	0.000	1.491983	2.900164
	age		1.000432	.0108671	0.04	0.968	.979358	1.02196
	tenure		1.030924	.0117703	2.67	0.008	1.00811	1.054253

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -279.03791  
Iteration 1: log likelihood = -260.07436  
Iteration 2: log likelihood = -258.56852  
Iteration 3: log likelihood = -256.56701  
Iteration 4: log likelihood = -255.97787  
Iteration 5: log likelihood = -255.92604  
Iteration 6: log likelihood = -255.91451  
Iteration 7: log likelihood = -255.912  
Iteration 8: log likelihood = -255.91149  
Iteration 9: log likelihood = -255.91137  
Iteration 10: log likelihood = -255.91134  
Iteration 11: log likelihood = -255.91134

Multinomial logistic regression	Number of obs	=	292
	LR chi2(18)	=	46.25
	Prob > chi2	=	0.0003
Log likelihood = -255.91134	Pseudo R2	=	0.0829

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1			(base outcome)				
2	Asian		(omitted)				
	Black		1.46e-07	.0006275	-0.00	0.997	0 .
	Latino		1.65e-07	.0004943	-0.01	0.996	0 .
	Other		6.25e-07	.0020622	-0.00	0.997	0 .
	AsianMale		(omitted)				
	BlackMale		7.23e+13	3.37e+17	0.01	0.995	0 .
	LatinoMale		8.48e+07	2.54e+11	0.01	0.995	0 .
	OtherMale		.4131446	1948.665	-0.00	1.000	0 .
	male		1.40888	.607606	0.79	0.427	.6050303 3.280731
	age		.965863	.0306429	-1.09	0.274	.9076333 1.027828
	tenure		1.033144	.0285908	1.18	0.239	.9785996 1.090728
3	Asian		(omitted)				
	Black		3.08e-07	.0006701	-0.01	0.994	0 .
	Latino		2.87e-07	.0004414	-0.01	0.992	0 .
	Other		2.808482	4.051459	0.72	0.474	.1661673 47.46764
	AsianMale		(omitted)				
	BlackMale		1.72e+13	4.88e+16	0.01	0.991	0 .
	LatinoMale		4855889	7.46e+09	0.01	0.992	0 .
	OtherMale		.284715	.5810951	-0.62	0.538	.0052135 15.54848
	male		3.822772	1.115423	4.60	0.000	2.157801 6.772442

age	1.00256	.0191059	0.13	0.893	.9658034	1.040715
tenure	.991405	.0168803	-0.51	0.612	.9588661	1.025048

-----> region = 15

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -274.13512  
Iteration 1: log likelihood = -254.0614  
Iteration 2: log likelihood = -252.07569  
Iteration 3: log likelihood = -252.01303  
Iteration 4: log likelihood = -252.00332  
Iteration 5: log likelihood = -252.001  
Iteration 6: log likelihood = -252.0005  
Iteration 7: log likelihood = -252.0004  
Iteration 8: log likelihood = -252.00038

Multinomial logistic regression	Number of obs	=	291
	LR chi2(14)	=	44.27
	Prob > chi2	=	0.0001
Log likelihood = -252.00038	Pseudo R2	=	0.0807

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
Asian	.0000128	.016334	-0.01	0.993	0 .
Black	4.005128	5951.347	0.00	0.999	0 .
Latino	4.977422	4.312881	1.85	0.064	.9108518 27.19952
Other	(omitted)				
AsianMale	(omitted)				
BlackMale	(omitted)				
LatinoMale	.3173834	.3208451	-1.14	0.256	.0437621 2.301812
OtherMale	(omitted)				
male	4.348599	2.660862	2.40	0.016	1.310715 14.42749
age	.9825783	.0357944	-0.48	0.629	.9148686 1.055299
tenure	1.050681	.0331348	1.57	0.117	.9877044 1.117673
3					
Asian	6.92e-06	.0041929	-0.02	0.984	0 .
Black	874600.6	5.34e+08	0.02	0.982	0 .
Latino	1.139829	.7519137	0.20	0.843	.3128424 4.152923
Other	(omitted)				
AsianMale	(omitted)				
BlackMale	(omitted)				
LatinoMale	.3423972	.2689755	-1.36	0.172	.0734269 1.596632
OtherMale	(omitted)				
male	3.957885	1.155425	4.71	0.000	2.233429 7.013815
age	1.001208	.019485	0.06	0.951	.9637369 1.040135
tenure	1.005878	.0179732	0.33	0.743	.9712614 1.041729

-----> region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -395.99946  
Iteration 1: log likelihood = -373.58371  
Iteration 2: log likelihood = -372.71862  
Iteration 3: log likelihood = -372.51763

```

Iteration 4: log likelihood = -372.47509
Iteration 5: log likelihood = -372.46561
Iteration 6: log likelihood = -372.46387
Iteration 7: log likelihood = -372.46369
Iteration 8: log likelihood = -372.46366
Iteration 9: log likelihood = -372.46365

```

```

Multinomial logistic regression      Number of obs   =      381
                                     LR chi2(18)        =      47.07
                                     Prob > chi2         =      0.0002
Log likelihood = -372.46365          Pseudo R2        =      0.0594

```

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1							
	Asian	(omitted)					
	Black		1.433333	1.395591	0.37	0.712	.2125978 9.663515
	Latino		9.18e-08	.0001613	-0.01	0.993	0 .
	Other		7715863	1.43e+10	0.01	0.993	0 .
	AsianMale	(omitted)					
	BlackMale		2.434564	20114.39	0.00	1.000	0 .
	LatinoMale		2.98e+07	5.24e+10	0.01	0.992	0 .
	OtherMale		.7287364	1819.881	-0.00	1.000	0 .
	male		.4191603	.1055909	-3.45	0.001	.2558313 .6867627
	age		.9640439	.0168003	-2.10	0.036	.9316719 .9975408
	tenure		.9848039	.0157501	-0.96	0.338	.9544131 1.016162
2							
	Asian	(omitted)					
	Black		7.85e-07	.000961	-0.01	0.991	0 .
	Latino		1.74e-07	.0004236	-0.01	0.995	0 .
	Other		1.414574	4912.986	0.00	1.000	0 .
	AsianMale	(omitted)					
	BlackMale		3.26e+14	1.83e+18	0.01	0.995	0 .
	LatinoMale		1.22e+07	2.97e+10	0.01	0.995	0 .
	OtherMale		5686853	2.19e+10	0.00	0.997	0 .
	male		.8687416	.2777191	-0.44	0.660	.4642786 1.625558
	age		.959103	.0204735	-1.96	0.050	.9198035 1.000082
	tenure		1.007023	.0189913	0.37	0.711	.9704805 1.044942
3		(base outcome)					

```

-----> region = 17

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -402.67234
Iteration 1: log likelihood = -375.46578
Iteration 2: log likelihood = -374.38031
Iteration 3: log likelihood = -374.17127
Iteration 4: log likelihood = -374.11985
Iteration 5: log likelihood = -374.10996
Iteration 6: log likelihood = -374.10823
Iteration 7: log likelihood = -374.10805
Iteration 8: log likelihood = -374.10801
Iteration 9: log likelihood = -374.108

```

```

Multinomial logistic regression      Number of obs   =      405
                                     LR chi2(18)        =      57.13
                                     Prob > chi2         =      0.0000
Log likelihood = -374.108          Pseudo R2        =      0.0709

```

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
--	-------	--	-----	-----------	---	------	----------------------

1							
	Asian	(omitted)					
	Black	.235394	.2049625	-1.66	0.097	.0427205	1.297043
	Latino	.2854721	.2177271	-1.64	0.100	.0640265	1.272822
	Other	7.08e-08	.0001683	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1.70e-06	.0023325	-0.01	0.992	0	.
	LatinoMale	2.913765	2.76365	1.13	0.260	.4540527	18.69833
	OtherMale	5.14e+07	1.22e+11	0.01	0.994	0	.
	male	.2848108	.0708683	-5.05	0.000	.1748861	.4638286
	age	.9889089	.0159058	-0.69	0.488	.9582205	1.02058
	tenure	.985933	.0144898	-0.96	0.335	.9579387	1.014745

2							
	Asian	(omitted)					
	Black	1.091374	1.277234	0.07	0.940	.1101067	10.81767
	Latino	4.54e-07	.0005038	-0.01	0.990	0	.
	Other	1.59e-07	.0006502	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	2.86e-07	.0006432	-0.01	0.995	0	.
	LatinoMale	3862514	4.29e+09	0.01	0.989	0	.
	OtherMale	8.774776	39562.16	0.00	1.000	0	.
	male	1.0279	.3981655	0.07	0.943	.4810953	2.196195
	age	.9500637	.0218223	-2.23	0.026	.9082412	.993812
	tenure	.9949803	.0213333	-0.23	0.814	.9540342	1.037684

3 | (base outcome)

-----> region = 18

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -425.52761  
Iteration 1: log likelihood = -411.03922  
Iteration 2: log likelihood = -410.66745  
Iteration 3: log likelihood = -410.66419  
Iteration 4: log likelihood = -410.66419

Multinomial logistic regression	Number of obs	=	418
	LR chi2(14)	=	29.73
	Prob > chi2	=	0.0083
Log likelihood = -410.66419	Pseudo R2	=	0.0349

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	(omitted)					
	Black	1.399035	1.671857	0.28	0.779	.1344734	14.55529
	Latino	2.378144	1.52455	1.35	0.177	.6769567	8.354404
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	5.672064	9.534635	1.03	0.302	.2103232	152.9661
	LatinoMale	.475918	.3863909	-0.91	0.360	.096929	2.33674
	OtherMale	(omitted)					
	male	1.762669	.6087391	1.64	0.101	.895793	3.468438
	age	.9452127	.0226792	-2.35	0.019	.9017912	.9907249
	tenure	1.023707	.0201198	1.19	0.233	.9850227	1.06391
3							

Asian		(omitted)					
Black		1.253061	1.184063	0.24	0.811	.1966262	7.985524
Latino		1.851313	1.042126	1.09	0.274	.6142259	5.579964
Other		(omitted)					
AsianMale		(omitted)					
BlackMale		2.083222	3.129959	0.49	0.625	.1096072	39.59427
LatinoMale		.7618749	.5134662	-0.40	0.687	.2033367	2.854642
OtherMale		(omitted)					
male		1.949194	.5040373	2.58	0.010	1.174208	3.235677
age		.988408	.0174211	-0.66	0.508	.9548462	1.023149
tenure		1.036406	.0145414	2.55	0.011	1.008294	1.065302

-----> region = 19

```

Iteration 0: log likelihood = -652.09841
Iteration 1: log likelihood = -611.97685
Iteration 2: log likelihood = -610.99027
Iteration 3: log likelihood = -610.81378
Iteration 4: log likelihood = -610.77491
Iteration 5: log likelihood = -610.76557
Iteration 6: log likelihood = -610.76361
Iteration 7: log likelihood = -610.7632
Iteration 8: log likelihood = -610.76311
Iteration 9: log likelihood = -610.76308

```

Multinomial logistic regression	Number of obs	=	612
	LR chi2(22)	=	82.67
	Prob > chi2	=	0.0000
Log likelihood = -610.76308	Pseudo R2	=	0.0634

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
Asian		5.76e-07	.0009466	-0.01	0.993	0 .
Black		.5749414	.4133273	-0.77	0.441	.140505 2.352639
Latino		1.2997	.369608	0.92	0.357	.7443532 2.269381
Other		2.651429	3.800152	0.68	0.496	.1597711 44.00092
AsianMale		8.39e+12	1.73e+16	0.01	0.988	0 .
BlackMale		1.366084	1.531792	0.28	0.781	.1517133 12.30073
LatinoMale		2.079708	.8389549	1.82	0.070	.9432511 4.585401
OtherMale		7.35e-07	.0005942	-0.02	0.986	0 .
male		1.039301	.2745359	0.15	0.884	.6192886 1.744175
age		.9759608	.0134489	-1.77	0.077	.9499542 1.002679
tenure		1.049459	.0136014	3.72	0.000	1.023136 1.076459
-----						
3						
Asian		7.81e-07	.0014988	-0.01	0.994	0 .
Black		.4508546	.3790125	-0.95	0.343	.0867902 2.342082
Latino		2.036949	.653462	2.22	0.027	1.086203 3.819876
Other		5.29e-06	.0055913	-0.01	0.991	0 .
AsianMale		1360441	3.87e+09	0.00	0.996	0 .
BlackMale		3.022237	3.452594	0.97	0.333	.3220465 28.36211
LatinoMale		1.001595	.4507075	0.00	0.997	.4146293 2.419492
OtherMale		284767.2	3.01e+08	0.01	0.991	0 .
male		1.539612	.4631381	1.43	0.151	.8538021 2.776293
age		.9829168	.0155987	-1.09	0.278	.9528146 1.01397
tenure		1.093186	.0162999	5.98	0.000	1.061701 1.125604

-----> region = 20

```

Iteration 0: log likelihood = -1813.3241
Iteration 1: log likelihood = -1712.8153
Iteration 2: log likelihood = -1709.6268
Iteration 3: log likelihood = -1709.1672
Iteration 4: log likelihood = -1709.0899
Iteration 5: log likelihood = -1709.0725
Iteration 6: log likelihood = -1709.0681
Iteration 7: log likelihood = -1709.0673
Iteration 8: log likelihood = -1709.0672
Iteration 9: log likelihood = -1709.0671
Iteration 10: log likelihood = -1709.0671

```

```

Multinomial logistic regression      Number of obs   =      1867
                                     LR chi2(22)        =      208.51
                                     Prob > chi2         =      0.0000
Log likelihood = -1709.0671          Pseudo R2        =      0.0575

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	3.32e-07	.000476	-0.01	0.992	0	.
	Black	.9928674	.3291329	-0.02	0.983	.5184659	1.901351
	Latino	.6991746	.1383147	-1.81	0.070	.4744561	1.030327
	Other	1.63e-07	.0003238	-0.01	0.994	0	.
	AsianMale	.8445259	3027.277	-0.00	1.000	0	.
	BlackMale	1.795059	.8442848	1.24	0.214	.7140461	4.512647
	LatinoMale	2.307946	.6406575	3.01	0.003	1.339499	3.976571
	OtherMale	1.26e+07	2.50e+10	0.01	0.993	0	.
	male	1.645342	.2579243	3.18	0.001	1.210101	2.237128
	age	.9649224	.0085385	-4.04	0.000	.9483314	.9818036
	tenure	1.019134	.0090345	2.14	0.033	1.00158	1.036996
3							
	Asian	1.470186	1.268549	0.45	0.655	.2709651	7.976846
	Black	1.119157	.3539032	0.36	0.722	.6021732	2.079987
	Latino	.9649786	.1837422	-0.19	0.851	.6644156	1.401508
	Other	2.55e-07	.0005062	-0.01	0.994	0	.
	AsianMale	1.583666	2.647601	0.27	0.783	.0597875	41.94857
	BlackMale	.3825827	.2110623	-1.74	0.082	.1297587	1.128013
	LatinoMale	1.082328	.2969988	0.29	0.773	.6320955	1.853253
	OtherMale	1.2501	3200.975	0.00	1.000	0	.
	male	2.186584	.3371892	5.07	0.000	1.616232	2.958206
	age	.989062	.0088104	-1.23	0.217	.9719438	1.006482
	tenure	1.076755	.0092322	8.63	0.000	1.058811	1.095002

### 1992-93

```
.char ethnicity[omit] "White"
```

```
. mlogit admin Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale male
tenure age if validcert1993a==1, rrr iter(20)
```

```

Iteration 0: log likelihood = -20293.243
Iteration 1: log likelihood = -19715.148
Iteration 2: log likelihood = -19710.074
Iteration 3: log likelihood = -19710.071
Iteration 4: log likelihood = -19710.071

```

```

Multinomial logistic regression      Number of obs   =      19719
                                     LR chi2(22)        =     1166.34
                                     Prob > chi2         =      0.0000
Log likelihood = -19710.071          Pseudo R2        =      0.0287

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	Asian	.551651	.2684988	-1.22	0.222	.2125048	1.432056
	Black	1.127693	.0833725	1.63	0.104	.9755731	1.303532
	Latino	1.20996	.0851686	2.71	0.007	1.054035	1.38895
	Other	.4380633	.1925379	-1.88	0.060	.1851041	1.036711
	AsianMale	2.939393	1.732362	1.83	0.067	.9259617	9.330871
	BlackMale	1.79322	.2075284	5.05	0.000	1.429304	2.249794
	LatinoMale	1.517457	.1530586	4.13	0.000	1.245259	1.849154
	OtherMale	1.690746	.950875	0.93	0.350	.561517	5.090891
	male	1.429909	.0634265	8.06	0.000	1.310845	1.559786
	tenure	1.006003	.0025025	2.41	0.016	1.00111	1.01092
	age	.9716549	.002549	-10.96	0.000	.9666717	.9766637
3							
	Asian	.6607302	.2564677	-1.07	0.286	.3087618	1.41392
	Black	.7369161	.0582495	-3.86	0.000	.6311534	.8604016
	Latino	1.235682	.0842474	3.10	0.002	1.081117	1.412344
	Other	.3378834	.1607838	-2.28	0.023	.132959	.8586499
	AsianMale	.8899737	.4512888	-0.23	0.818	.3294193	2.404393
	BlackMale	1.076934	.1332748	0.60	0.549	.8449864	1.372552
	LatinoMale	.8384754	.082734	-1.79	0.074	.6910359	1.017373
	OtherMale	1.942459	1.101367	1.17	0.242	.6393276	5.901741
	male	2.049662	.0821924	17.90	0.000	1.894736	2.217256
	tenure	1.027092	.0022618	12.14	0.000	1.022669	1.031535
	age	1.00924	.0024581	3.78	0.000	1.004434	1.014069

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=19719)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
Asian	2.385	2	0.303
Black	22.850	2	0.000
Latino	13.120	2	0.001
Other	9.313	2	0.009
AsianMale	4.240	2	0.120
BlackMale	27.049	2	0.000
LatinoMale	29.816	2	0.000
OtherMale	2.014	2	0.365
male	331.994	2	0.000
tenure	152.024	2	0.000
age	180.964	2	0.000

\*\*\*\* Wald tests for independent variables (N=19719)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
Asian	2.243	2	0.326
Black	22.013	2	0.000
Latino	13.167	2	0.001
Other	7.847	2	0.020
AsianMale	3.851	2	0.146
BlackMale	27.031	2	0.000
LatinoMale	29.709	2	0.000
OtherMale	1.927	2	0.382

male	327.169	2	0.000
tenure	149.886	2	0.000
age	176.182	2	0.000

\*\*\*\* Hausman tests of IIA assumption (N=19719)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-1.823	11	---	---
3	-5.171	12	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=19719)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-4815.298	-4812.957	4.682	12	0.968	for Ho
3	-4210.175	-4205.231	9.889	12	0.626	for Ho

\*\*\*\* Wald tests for combining alternatives (N=19719)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2- 3	538.144	11	0.000
2- 1	425.329	11	0.000
3- 1	692.342	11	0.000

\*\*\*\* LR tests for combining alternatives (N=19719)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2- 3	566.527	11	0.000
2- 1	432.993	11	0.000
3- 1	724.994	11	0.000

by region

. bysort region: mlogit admin Asian Black Latino Other AsianMale BlackMale LatinoMale  
OtherMale male tenure age if validcert1993a==1, rrr iter(20)

-----> region = 1

note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1293.7549  
Iteration 1: log likelihood = -1228.0739  
Iteration 2: log likelihood = -1227.1403  
Iteration 3: log likelihood = -1226.9816  
Iteration 4: log likelihood = -1226.9555  
Iteration 5: log likelihood = -1226.9529  
Iteration 6: log likelihood = -1226.9523



Iteration 7: log likelihood = -1226.9522  
 Iteration 8: log likelihood = -1226.9522  
 Iteration 9: log likelihood = -1226.9522

Multinomial logistic regression      Number of obs    =      1221  
    LR chi2(20)       =      133.61  
    Prob > chi2       =      0.0000  
 Log likelihood = -1226.9522          Pseudo R2        =      0.0516

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	7.90e-07	.0014315	-0.01	0.994	0	.
	Black	4.21e+07	1.31e+11	0.01	0.995	0	.
	Latino	1.273016	.3311485	0.93	0.353	.7645601	2.119611
	Other	4.34e-07	.0011222	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	4.46e-08	.0001385	-0.01	0.996	0	.
	LatinoMale	1.366907	.4792595	0.89	0.373	.6875335	2.717589
	OtherMale	5780942	1.49e+10	0.01	0.995	0	.
	male	1.808294	.5594837	1.91	0.056	.9860718	3.316116
	tenure	1.025692	.0102945	2.53	0.011	1.005713	1.046069
	age	.944906	.0101946	-5.25	0.000	.9251347	.9650998
3							
	Asian	1.98e-07	.0003617	-0.01	0.993	0	.
	Black	1.824156	9399.817	0.00	1.000	0	.
	Latino	.9018492	.2037913	-0.46	0.648	.5791454	1.404366
	Other	2.26e-07	.0005839	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	7.36e-07	.0039547	-0.00	0.998	0	.
	LatinoMale	2.807736	.9830306	2.95	0.003	1.413636	5.576667
	OtherMale	1.90e+07	4.91e+10	0.01	0.995	0	.
	male	.6990935	.215837	-1.16	0.246	.3817128	1.280365
	tenure	1.054788	.0099869	5.63	0.000	1.035395	1.074545
	age	.9906821	.0104549	-0.89	0.375	.9704014	1.011387

-----> region = 2

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -812.84779  
 Iteration 1: log likelihood = -784.24804  
 Iteration 2: log likelihood = -783.60314  
 Iteration 3: log likelihood = -782.94859  
 Iteration 4: log likelihood = -782.90821  
 Iteration 5: log likelihood = -782.90159  
 Iteration 6: log likelihood = -782.90088  
 Iteration 7: log likelihood = -782.90074  
 Iteration 8: log likelihood = -782.90072

Multinomial logistic regression      Number of obs    =      865  
    LR chi2(18)       =      59.89  
    Prob > chi2       =      0.0000  
 Log likelihood = -782.90072          Pseudo R2        =      0.0368

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						

```

      Asian | (omitted)
      Black | 2.330268 1.496482 1.32 0.188 .6618658 8.204303
      Latino | .8198668 .2295086 -0.71 0.478 .473656 1.419134
      Other | 7.248265 9.055659 1.59 0.113 .6262976 83.88558
      AsianMale | (omitted)
      BlackMale | 1.387109 1.356756 0.33 0.738 .2039555 9.43378
      LatinoMale | 1.73905 .6997991 1.38 0.169 .7902888 3.826822
      OtherMale | 4.33e-07 .0002481 -0.03 0.980 0 .
      male | 1.564419 .4298357 1.63 0.103 .9130178 2.680569
      tenure | 1.01919 .0144266 1.34 0.179 .9913035 1.047862
      age | .9709381 .0144458 -1.98 0.047 .9430336 .9996682
-----+-----
3 |
      Asian | (omitted)
      Black | .9376042 .7604193 -0.08 0.937 .1912823 4.595833
      Latino | 1.139622 .2777548 0.54 0.592 .7068077 1.83747
      Other | 3.52e-06 .0044091 -0.01 0.992 0 .
      AsianMale | (omitted)
      BlackMale | 1.924681 2.09818 0.60 0.548 .2272094 16.30389
      LatinoMale | 1.209084 .4176718 0.55 0.583 .6143468 2.379576
      OtherMale | .9118975 1221.356 -0.00 1.000 0 .
      male | 1.997576 .4770879 2.90 0.004 1.250858 3.190059
      tenure | 1.016722 .0117297 1.44 0.151 .9939906 1.039974
      age | 1.01655 .0127401 1.31 0.190 .9918845 1.04183
-----+-----

```

```

-----> region = 3

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -337.96099
Iteration 1: log likelihood = -323.08751
Iteration 2: log likelihood = -322.14724
Iteration 3: log likelihood = -322.00021
Iteration 4: log likelihood = -321.96601
Iteration 5: log likelihood = -321.9589
Iteration 6: log likelihood = -321.95741
Iteration 7: log likelihood = -321.95707
Iteration 8: log likelihood = -321.95699
Iteration 9: log likelihood = -321.95697

```

```

Multinomial logistic regression      Number of obs   =      339
                                     LR chi2(16)        =      32.01
                                     Prob > chi2         =      0.0100
Log likelihood = -321.95697          Pseudo R2       =      0.0474

```

```

-----+-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1 |          |
  | (base outcome)
-----+-----
2 |          |
  | Asian | (omitted)
  | Black | 1.764168 1.574174 0.64 0.525 .3069091 10.14075
  | Latino | 4.61394 3.978864 1.77 0.076 .8512013 25.00988
  | Other | .7686698 2458.053 -0.00 1.000 0 .
  | AsianMale | (omitted)
  | BlackMale | .6192027 .9143782 -0.32 0.745 .0342658 11.18934
  | LatinoMale | .3535436 .409148 -0.90 0.369 .0365904 3.41601
  | OtherMale | (omitted)
  | male | 1.327813 .4853885 0.78 0.438 .6485986 2.718302
  | tenure | .9794533 .0202513 -1.00 0.315 .9405549 1.01996
  | age | .9937485 .0265596 -0.23 0.814 .9430325 1.047192
-----+-----
3 |          |

```

Asian		(omitted)					
Black		1.184269	1.033363	0.19	0.846	.2141475	6.54919
Latino		.8256319	.9703005	-0.16	0.870	.0824966	8.262986
Other		2755284	4.21e+09	0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		7.58e-07	.0004945	-0.02	0.983	0	.
LatinoMale		2.621319	3.403565	0.74	0.458	.2057339	33.39905
OtherMale		(omitted)					
male		2.393909	.6582866	3.17	0.002	1.396499	4.103688
tenure		.9891259	.0147517	-0.73	0.463	.9606315	1.018465
age		1.031087	.0210086	1.50	0.133	.9907226	1.073097

-----> region = 4

Iteration 0: log likelihood = -4028.5772  
Iteration 1: log likelihood = -3819.0789  
Iteration 2: log likelihood = -3814.2069  
Iteration 3: log likelihood = -3814.0288  
Iteration 4: log likelihood = -3813.9863  
Iteration 5: log likelihood = -3813.9776  
Iteration 6: log likelihood = -3813.9757  
Iteration 7: log likelihood = -3813.9753  
Iteration 8: log likelihood = -3813.9752  
Iteration 9: log likelihood = -3813.9752

Multinomial logistic regression	Number of obs	=	3912
	LR chi2(22)	=	429.20
	Prob > chi2	=	0.0000
Log likelihood = -3813.9752	Pseudo R2	=	0.0533

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)					
2							
	Asian		.6749979	.5546489	-0.48	0.632	.1348547 3.378614
	Black		.5149712	.0635266	-5.38	0.000	.4043702 .6558231
	Latino		1.280341	.2566576	1.23	0.218	.8643577 1.896522
	Other		.2569657	.2760902	-1.26	0.206	.0312841 2.110701
	AsianMale		1.58e-06	.000829	-0.03	0.980	0 .
	BlackMale		2.618912	.5165504	4.88	0.000	1.779226 3.854879
	LatinoMale		2.13883	.6984317	2.33	0.020	1.127762 4.056345
	OtherMale		2.563616	3.472388	0.70	0.487	.1802646 36.45822
	male		1.102333	.1021997	1.05	0.293	.9191705 1.321994
	tenure		1.02058	.0057058	3.64	0.000	1.009458 1.031825
	age		.9647947	.0053796	-6.43	0.000	.9543083 .9753963
3							
	Asian		1.031033	.8639456	0.04	0.971	.1995332 5.327579
	Black		.3189582	.048099	-7.58	0.000	.2373407 .4286426
	Latino		2.409716	.482197	4.40	0.000	1.627931 3.566941
	Other		.5356667	.5806734	-0.58	0.565	.0639993 4.483469
	AsianMale		.2336845	.323558	-1.05	0.294	.0154904 3.525314
	BlackMale		2.292682	.5335488	3.57	0.000	1.45296 3.617711
	LatinoMale		.8310216	.3021869	-0.51	0.611	.4074621 1.694874
	OtherMale		1.392801	1.914839	0.24	0.810	.0941126 20.6125
	male		1.289254	.1322809	2.48	0.013	1.054394 1.576428
	tenure		1.081486	.0066367	12.77	0.000	1.068557 1.094573
	age		.9969529	.0064075	-0.47	0.635	.9844731 1.009591

-----> region = 5

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -571.66916
Iteration 1: log likelihood = -515.49435
Iteration 2: log likelihood = -513.25062
Iteration 3: log likelihood = -513.05212
Iteration 4: log likelihood = -513.02096
Iteration 5: log likelihood = -513.01351
Iteration 6: log likelihood = -513.01192
Iteration 7: log likelihood = -513.01159
Iteration 8: log likelihood = -513.01152
Iteration 9: log likelihood = -513.0115
```

```
Multinomial logistic regression      Number of obs   =      621
                                     LR chi2(20)        =      117.32
                                     Prob > chi2         =      0.0000
Log likelihood = -513.0115           Pseudo R2        =      0.1026
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	1.74e+08	8.66e+11	0.00	0.997	0	.
	Black	2.049427	.8550763	1.72	0.085	.904666	4.642765
	Latino	8.2214	10.54171	1.64	0.100	.6660571	101.4799
	Other	1.58e-06	.003481	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	.817476	.4632552	-0.36	0.722	.2692204	2.48223
	LatinoMale	.0914969	.1639271	-1.33	0.182	.0027314	3.064996
	OtherMale	.1806247	488.8667	-0.00	0.999	0	.
	male	5.047287	1.491218	5.48	0.000	2.828596	9.006272
	tenure	1.051517	.0175157	3.02	0.003	1.017742	1.086414
	age	.940732	.0176983	-3.25	0.001	.9066757	.9760675
3							
	Asian	2.553161	25204.57	0.00	1.000	0	.
	Black	1.451209	.5684262	0.95	0.342	.6734755	3.127076
	Latino	5.09e-06	.0054649	-0.01	0.991	0	.
	Other	8.60e-07	.0016688	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.495444	.2796913	-1.24	0.213	.1638584	1.49803
	LatinoMale	.1147747	173.6505	-0.00	0.999	0	.
	OtherMale	.4838494	1139.273	-0.00	1.000	0	.
	male	4.336104	1.058013	6.01	0.000	2.687855	6.995094
	tenure	1.043774	.0138433	3.23	0.001	1.016991	1.071262
	age	1.018502	.0155878	1.20	0.231	.9884045	1.049517

-----> region = 6

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -683.55425
Iteration 1: log likelihood = -659.22975
Iteration 2: log likelihood = -658.44229
Iteration 3: log likelihood = -658.28957
Iteration 4: log likelihood = -658.25562
Iteration 5: log likelihood = -658.24723
Iteration 6: log likelihood = -658.24561
Iteration 7: log likelihood = -658.24535
Iteration 8: log likelihood = -658.24528
Iteration 9: log likelihood = -658.24527
```

```
Multinomial logistic regression      Number of obs   =      659
                                     LR chi2(20)        =      50.62
                                     Prob > chi2         =      0.0002
```

Log likelihood = -658.24527

Pseudo R2 = 0.0370

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	5.01e-07	.0010516	-0.01	0.994	0	.
	Black	.700127	.4095901	-0.61	0.542	.2224358	2.203682
	Latino	.768678	.8956938	-0.23	0.821	.0783233	7.543935
	Other	1.01e-06	.0021235	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	3.57158	2.82996	1.61	0.108	.7557994	16.87774
	LatinoMale	5.359191	9.146438	0.98	0.325	.1889594	151.9953
	OtherMale	795372.6	3.57e+09	0.00	0.998	0	.
	male	1.392944	.3019746	1.53	0.126	.9107571	2.130416
	tenure	1.007788	.0135881	0.58	0.565	.9815051	1.034776
	age	.9747866	.0137282	-1.81	0.070	.9482478	1.002068
3							
	Asian	9.06e-07	.0017049	-0.01	0.994	0	.
	Black	.6284397	.4105501	-0.71	0.477	.1746573	2.261208
	Latino	1.63e-06	.0013214	-0.02	0.987	0	.
	Other	5.13e-07	.0009644	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1.406311	1.227537	0.39	0.696	.2541482	7.781727
	LatinoMale	601401	4.86e+08	0.02	0.987	0	.
	OtherMale	1.08e+13	3.13e+16	0.01	0.992	0	.
	male	2.514214	.5001077	4.64	0.000	1.702503	3.712928
	tenure	.9993434	.0118347	-0.06	0.956	.9764149	1.02281
	age	1.022028	.0131743	1.69	0.091	.9965301	1.048178

-----> region = 7

Iteration 0: log likelihood = -971.46592  
 Iteration 1: log likelihood = -920.52062  
 Iteration 2: log likelihood = -911.82877  
 Iteration 3: log likelihood = -911.45247  
 Iteration 4: log likelihood = -911.37634  
 Iteration 5: log likelihood = -911.35775  
 Iteration 6: log likelihood = -911.35383  
 Iteration 7: log likelihood = -911.35301  
 Iteration 8: log likelihood = -911.35282  
 Iteration 9: log likelihood = -911.35277  
 Iteration 10: log likelihood = -911.35277

Multinomial logistic regression

Number of obs = 1039  
 LR chi2(22) = 120.23  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0619

Log likelihood = -911.35277

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	4.20e+09	7.00e+13	0.00	0.999	0	.
	Black	1.286962	.6155912	0.53	0.598	.5039774	3.286397
	Latino	1.85e-06	.0022367	-0.01	0.991	0	.
	Other	8.92e-07	.0017242	-0.01	0.994	0	.
	AsianMale	.0068013	114.4989	-0.00	1.000	0	.
	BlackMale	2.644911	1.5315	1.68	0.093	.8502257	8.227879
	LatinoMale	.3882057	888.739	-0.00	1.000	0	.

OtherMale		3951947	7.64e+09	0.01	0.994	0	.
male		3.748937	.9315542	5.32	0.000	2.303548	6.10125
tenure		1.010853	.0140275	0.78	0.437	.9837296	1.038723
age		.9700838	.0140796	-2.09	0.036	.9428771	.9980756
-----							
3							
Asian		1.132691	32155.97	0.00	1.000	0	.
Black		.8370311	.2657075	-0.56	0.575	.4492994	1.559363
Latino		.6127755	.676787	-0.44	0.657	.0703367	5.33852
Other		7.31e-07	.0008994	-0.01	0.991	0	.
AsianMale		5220043	1.49e+11	0.00	1.000	0	.
BlackMale		.9578056	.4475179	-0.09	0.926	.3833261	2.393241
LatinoMale		3.897043	6.458844	0.82	0.412	.1513487	100.3441
OtherMale		1764774	2.17e+09	0.01	0.991	0	.
male		2.629611	.4096864	6.21	0.000	1.937656	3.568671
tenure		1.009357	.0093885	1.00	0.317	.9911225	1.027927
age		1.007686	.0100172	0.77	0.441	.9882427	1.027512
-----							

-----> region = 8

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -294.34991  
Iteration 1: log likelihood = -272.24126  
Iteration 2: log likelihood = -270.6555  
Iteration 3: log likelihood = -270.46275  
Iteration 4: log likelihood = -270.43085  
Iteration 5: log likelihood = -270.42365  
Iteration 6: log likelihood = -270.42188  
Iteration 7: log likelihood = -270.42152  
Iteration 8: log likelihood = -270.42145  
Iteration 9: log likelihood = -270.42143

Multinomial logistic regression	Number of obs	=	307
	LR chi2(18)	=	47.86
	Prob > chi2	=	0.0002
Log likelihood = -270.42143	Pseudo R2	=	0.0813

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]		
1	(base outcome)						
2							
Asian	3.62e-07	.0010222	-0.01	0.996	0		.
Black	3.503132	4.255445	1.03	0.302	.3239332	37.88414	
Latino	2.78e-06	.0055425	-0.01	0.995	0		.
Other	2.45e-06	.0048929	-0.01	0.995	0		.
AsianMale	(omitted)						
BlackMale	.8811255	1.238198	-0.09	0.928	.0560907	13.84155	
LatinoMale	.1127487	390.0413	-0.00	0.999	0		.
OtherMale	(omitted)						
male	7.630241	4.305855	3.60	0.000	2.524602	23.06129	
tenure	1.016699	.0250892	0.67	0.502	.9686955	1.067082	
age	1.008589	.0260754	0.33	0.741	.9587558	1.061013	
3							
Asian	3.79e-07	.0006103	-0.01	0.993	0		.
Black	1.618029	1.250964	0.62	0.534	.3555394	7.363508	
Latino	1.62e-06	.0018455	-0.01	0.991	0		.
Other	1.06e-06	.0012009	-0.01	0.990	0		.
AsianMale	(omitted)						
BlackMale	.9235449	.9326835	-0.08	0.937	.127595	6.684706	
LatinoMale	.2659124	523.9912	-0.00	0.999	0		.
OtherMale	(omitted)						

male		3.405279	.9716863	4.29	0.000	1.946547	5.957178
tenure		1.032157	.0177999	1.84	0.066	.9978525	1.06764
age		.989381	.0176199	-0.60	0.549	.9554424	1.024525

-----> region = 9

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -218.72699  
Iteration 1: log likelihood = -202.29017  
Iteration 2: log likelihood = -201.8171  
Iteration 3: log likelihood = -201.73043  
Iteration 4: log likelihood = -201.71224  
Iteration 5: log likelihood = -201.70918  
Iteration 6: log likelihood = -201.70846  
Iteration 7: log likelihood = -201.7083  
Iteration 8: log likelihood = -201.70826  
Iteration 9: log likelihood = -201.70826

Multinomial logistic regression	Number of obs	=	221
	LR chi2(16)	=	34.04
	Prob > chi2	=	0.0054
	Pseudo R2	=	0.0778

Log likelihood = -201.70826

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	1.67e-07	.0008939	-0.00	0.998	0	.
	Black	2.807724	4.197322	0.69	0.490	.1499259	52.58142
	Latino	.3401698	2227.883	-0.00	1.000	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	2.80e-08	.0000833	-0.01	0.995	0	.
	LatinoMale	7.38e+09	2.22e+14	0.00	0.999	0	.
	OtherMale	(omitted)					
	male	1.715926	.8002927	1.16	0.247	.6878686	4.28047
	tenure	1.039225	.0305446	1.31	0.191	.9810501	1.100849
	age	1.025923	.0363723	0.72	0.470	.9570547	1.099746
3							
	Asian	2.33e-07	.0007054	-0.01	0.996	0	.
	Black	2.708507	3.507507	0.77	0.442	.2140065	34.27938
	Latino	.6601108	2.04e+10	0.01	0.996	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	2.28e-08	.0000386	-0.01	0.992	0	.
	LatinoMale	9.02e-08	.0038625	-0.00	1.000	0	.
	OtherMale	(omitted)					
	male	2.749708	.9126798	3.05	0.002	1.434686	5.27007
	tenure	1.005202	.0201715	0.26	0.796	.9664343	1.045526
	age	1.060549	.0256001	2.44	0.015	1.011542	1.11193

-----> region = 10

Iteration 0: log likelihood = -2713.0213  
Iteration 1: log likelihood = -2607.7513  
Iteration 2: log likelihood = -2606.2205  
Iteration 3: log likelihood = -2605.9578  
Iteration 4: log likelihood = -2605.9004

```

Iteration 5: log likelihood = -2605.8867
Iteration 6: log likelihood = -2605.8838
Iteration 7: log likelihood = -2605.8832
Iteration 8: log likelihood = -2605.8831
Iteration 9: log likelihood = -2605.883
Iteration 10: log likelihood = -2605.883

```

```

Multinomial logistic regression
Log likelihood = -2605.883
Number of obs = 2651
LR chi2(22) = 214.28
Prob > chi2 = 0.0000
Pseudo R2 = 0.0395

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	.3631945	.3883837	-0.95	0.344	.0446584	2.953762
	Black	1.442045	.231266	2.28	0.022	1.053098	1.974646
	Latino	2.38186	.6626176	3.12	0.002	1.380759	4.108799
	Other	.6566632	.5091652	-0.54	0.588	.14366	3.001578
	AsianMale	2.865885	3.301887	0.91	0.361	.2996123	27.41308
	BlackMale	1.066208	.2851732	0.24	0.811	.631212	1.800978
	LatinoMale	.8150418	.3500486	-0.48	0.634	.3512377	1.891292
	OtherMale	.4395535	.583565	-0.62	0.536	.0325784	5.930532
	male	1.893183	.2267132	5.33	0.000	1.497129	2.39401
	tenure	1.032916	.0072859	4.59	0.000	1.018734	1.047295
	age	.9687848	.0067974	-4.52	0.000	.9555533	.9821995
3							
	Asian	.7389403	.4415959	-0.51	0.613	.2290473	2.38393
	Black	.8507456	.1479999	-0.93	0.353	.6049515	1.196407
	Latino	2.326537	.629773	3.12	0.002	1.368665	3.954783
	Other	3.84e-07	.000313	-0.02	0.986	0	.
	AsianMale	.8330715	.5937934	-0.26	0.798	.206047	3.368203
	BlackMale	1.140786	.3228435	0.47	0.642	.6551084	1.986529
	LatinoMale	.6745708	.2884772	-0.92	0.357	.2917527	1.559697
	OtherMale	2808477	2.29e+09	0.02	0.985	0	.
	male	1.923829	.2150974	5.85	0.000	1.545241	2.395172
	tenure	1.053934	.0068924	8.03	0.000	1.040511	1.06753
	age	1.00058	.0066705	0.09	0.931	.9875907	1.013739

-----> region = 11

```

Iteration 0: log likelihood = -1883.7931
Iteration 1: log likelihood = -1757.7272
Iteration 2: log likelihood = -1753.5068
Iteration 3: log likelihood = -1753.3304
Iteration 4: log likelihood = -1753.2896
Iteration 5: log likelihood = -1753.2808
Iteration 6: log likelihood = -1753.2788
Iteration 7: log likelihood = -1753.2784
Iteration 8: log likelihood = -1753.2784
Iteration 9: log likelihood = -1753.2784

```

```

Multinomial logistic regression
Log likelihood = -1753.2784
Number of obs = 1782
LR chi2(22) = 261.03
Prob > chi2 = 0.0000
Pseudo R2 = 0.0693

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					



2							
	Asian	2.84e-07	.000455	-0.01	0.992	0	.
	Black	1.717036	.4715758	1.97	0.049	1.002307	2.941425
	Latino	1.666111	.5818015	1.46	0.144	.8403625	3.303248
	Other	.9661161	1.202898	-0.03	0.978	.0841802	11.08788
	AsianMale	1.32e+14	5.66e+17	0.01	0.994	0	.
	BlackMale	1.343951	.5695535	0.70	0.485	.585669	3.084
	LatinoMale	1.195013	.6565977	0.32	0.746	.4070816	3.508034
	OtherMale	.693187	1.21788	-0.21	0.835	.0221488	21.69455
	male	1.119826	.1531211	0.83	0.408	.856565	1.463999
	tenure	.9173739	.0081133	-9.75	0.000	.9016092	.9334143
	age	.9797568	.0081274	-2.47	0.014	.9639561	.9958164
3							
	Asian	4.70e-07	.0007561	-0.01	0.993	0	.
	Black	1.655665	.4733062	1.76	0.078	.9454507	2.899386
	Latino	1.678148	.635049	1.37	0.171	.799312	3.523255
	Other	7.94e-07	.0009138	-0.01	0.990	0	.
	AsianMale	1010140	6.85e+09	0.00	0.998	0	.
	BlackMale	.3542065	.1731137	-2.12	0.034	.1359074	.923145
	LatinoMale	.6030018	.3591971	-0.85	0.396	.1876176	1.938044
	OtherMale	1203214	1.38e+09	0.01	0.990	0	.
	male	1.860319	.2332692	4.95	0.000	1.454969	2.3786
	tenure	.9578631	.0067091	-6.15	0.000	.9448034	.9711033
	age	1.042624	.0083829	5.19	0.000	1.026322	1.059184

-----> region = 12

note: AsianMale omitted because of collinearity

note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -690.34259

Iteration 1: log likelihood = -659.62891

Iteration 2: log likelihood = -657.3714

Iteration 3: log likelihood = -656.25536

Iteration 4: log likelihood = -656.16597

Iteration 5: log likelihood = -656.15101

Iteration 6: log likelihood = -656.14752

Iteration 7: log likelihood = -656.14675

Iteration 8: log likelihood = -656.1466

Iteration 9: log likelihood = -656.14656

Iteration 10: log likelihood = -656.14655

Multinomial logistic regression

Number of obs = 687

LR chi2(18) = 68.39

Prob > chi2 = 0.0000

Log likelihood = -656.14655

Pseudo R2 = 0.0495

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	Asian	6.380289	7.933526	1.49	0.136	.5577244 72.98961
	Black	1.940872	.8697708	1.48	0.139	.8063887 4.671424
	Latino	7.58e-07	.0007682	-0.01	0.989	0 .
	Other	1.94e+08	1.69e+12	0.00	0.998	0 .
	AsianMale	(omitted)				
	BlackMale	2.118565	1.425393	1.12	0.264	.5666884 7.920256
	LatinoMale	.6190887	6.27e+09	0.02	0.988	0 .
	OtherMale	(omitted)				
	male	1.221074	.2993604	0.81	0.415	.755197 1.974349
	tenure	1.005884	.0150586	0.39	0.695	.9767985 1.035836
	age	.9801191	.0151416	-1.30	0.194	.9508868 1.01025

3							
	Asian	6.64e-08	.0002533	-0.00	0.997	0	.
	Black	1.714289	.7006099	1.32	0.187	.7694985	3.819092
	Latino	2.27264	1.570759	1.19	0.235	.5864266	8.807398
	Other	.8484396	11774.89	-0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.6251861	.4029873	-0.73	0.466	.1767401	2.211482
	LatinoMale	7.21e-08	.0001063	-0.01	0.991	0	.
	OtherMale	(omitted)					
	male	2.715471	.5235775	5.18	0.000	1.860898	3.962486
	tenure	1.01556	.0114633	1.37	0.171	.9933387	1.038278
	age	1.012243	.0120855	1.02	0.308	.988831	1.03621

-----> region = 13

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1105.2241  
Iteration 1: log likelihood = -1084.0481  
Iteration 2: log likelihood = -1083.5849  
Iteration 3: log likelihood = -1083.4899  
Iteration 4: log likelihood = -1083.4682  
Iteration 5: log likelihood = -1083.4637  
Iteration 6: log likelihood = -1083.463  
Iteration 7: log likelihood = -1083.4629  
Iteration 8: log likelihood = -1083.4629

Multinomial logistic regression	Number of obs	=	1031
	LR chi2(18)	=	43.52
	Prob > chi2	=	0.0007
Log likelihood = -1083.4629	Pseudo R2	=	0.0197

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	Asian	(omitted)				
	Black	1.870269	.674126	1.74	0.082	.9227713 3.790653
	Latino	2.102091	.6476108	2.41	0.016	1.149248 3.844936
	Other	4.40e-06	.0038672	-0.01	0.989	0 .
	AsianMale	(omitted)				
	BlackMale	1.352766	.8376953	0.49	0.626	.4019014 4.553298
	LatinoMale	.7598421	.355185	-0.59	0.557	.3039714 1.899389
	OtherMale	212775.5	1.87e+08	0.01	0.989	0 .
	male	1.456169	.2666453	2.05	0.040	1.017055 2.08487
	tenure	1.024396	.0122725	2.01	0.044	1.000622 1.048734
	age	.9726505	.0113922	-2.37	0.018	.9505766 .9952369
3						
	Asian	(omitted)				
	Black	1.386359	.5014782	0.90	0.366	.6823019 2.816924
	Latino	1.603349	.4911623	1.54	0.123	.8795794 2.922679
	Other	1.801116	2.567435	0.41	0.680	.1101996 29.43765
	AsianMale	(omitted)				
	BlackMale	1.718552	1.031835	0.90	0.367	.5297776 5.574834
	LatinoMale	.6162818	.2912129	-1.02	0.306	.2440952 1.555963
	OtherMale	7.21e-07	.0004671	-0.02	0.983	0 .
	male	1.510806	.2490695	2.50	0.012	1.093656 2.087068
	tenure	1.021958	.0110195	2.01	0.044	1.000586 1.043785
	age	1.00644	.0106852	0.60	0.545	.9857135 1.027602

-----> region = 14

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -276.77506  
 Iteration 1: log likelihood = -259.38687  
 Iteration 2: log likelihood = -257.21454  
 Iteration 3: log likelihood = -256.5753  
 Iteration 4: log likelihood = -256.43785  
 Iteration 5: log likelihood = -256.40397  
 Iteration 6: log likelihood = -256.39741  
 Iteration 7: log likelihood = -256.3963  
 Iteration 8: log likelihood = -256.39604  
 Iteration 9: log likelihood = -256.39598  
 Iteration 10: log likelihood = -256.39597

Multinomial logistic regression	Number of obs	=	286
	LR chi2(18)	=	40.76
	Prob > chi2	=	0.0016
Log likelihood = -256.39597	Pseudo R2	=	0.0736

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
Asian	(omitted)					
Black	6.494237	9.740987	1.25	0.212	.3433793	122.8237
Latino	1.66e-07	.0006985	-0.00	0.997	0	.
Other	6.06e-07	.0024901	-0.00	0.997	0	.
AsianMale	(omitted)					
BlackMale	.1003097	530.7418	-0.00	1.000	0	.
LatinoMale	1.78e+14	8.36e+17	0.01	0.994	0	.
OtherMale	.2440592	1405.692	-0.00	1.000	0	.
male	1.688023	.7142765	1.24	0.216	.7365425	3.868647
tenure	1.029223	.0291674	1.02	0.309	.9736146	1.088008
age	.9316668	.0298298	-2.21	0.027	.8749981	.9920055
3						
Asian	(omitted)					
Black	4.33e-08	.0003321	-0.00	0.998	0	.
Latino	1.59e-07	.0003673	-0.01	0.995	0	.
Other	2.619669	3.777086	0.67	0.504	.1552276	44.21036
AsianMale	(omitted)					
BlackMale	3.31e+14	2.65e+18	0.00	0.997	0	.
LatinoMale	6.70e+13	2.10e+17	0.01	0.992	0	.
OtherMale	.3380447	.6916386	-0.53	0.596	.0061294	18.64363
male	2.779554	.7886317	3.60	0.000	1.593922	4.847113
tenure	.9945262	.0171545	-0.32	0.750	.961466	1.028723
age	.9894551	.0196884	-0.53	0.594	.9516094	1.028806

-----> region = 15

note: Other omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: BlackMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -279.27809  
 Iteration 1: log likelihood = -261.94179  
 Iteration 2: log likelihood = -259.93432  
 Iteration 3: log likelihood = -259.87313  
 Iteration 4: log likelihood = -259.8633  
 Iteration 5: log likelihood = -259.86103

```

Iteration 6:  log likelihood = -259.8605
Iteration 7:  log likelihood = -259.86039
Iteration 8:  log likelihood = -259.86037

```

```

Multinomial logistic regression      Number of obs   =      298
                                     LR chi2(14)        =      38.84
                                     Prob > chi2         =      0.0004
Log likelihood = -259.86037          Pseudo R2        =      0.0695

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	Asian	3.46e-06	.0070002	-0.01	0.995	0	.
	Black	2.683234	6890.184	0.00	1.000	0	.
	Latino	7.281147	5.476246	2.64	0.008	1.667247	31.79799
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.29734	.2725137	-1.32	0.186	.0493313	1.79219
	OtherMale	(omitted)					
	male	2.478906	1.411639	1.59	0.111	.8119597	7.568081
	tenure	1.023673	.0296894	0.81	0.420	.967106	1.083549
	age	.9908	.0317982	-0.29	0.773	.9303964	1.055125
3							
	Asian	4.43e-06	.0044645	-0.01	0.990	0	.
	Black	2939203	3.02e+09	0.01	0.988	0	.
	Latino	1.799505	1.132625	0.93	0.351	.5240815	6.178843
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.2920756	.2188513	-1.64	0.100	.0672501	1.268522
	OtherMale	(omitted)					
	male	3.670855	1.097485	4.35	0.000	2.04306	6.595583
	tenure	1.00649	.0169726	0.38	0.701	.9737675	1.040311
	age	1.019377	.018616	1.05	0.293	.9835353	1.056524

```

-----> region = 16

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0:  log likelihood = -421.18555
Iteration 1:  log likelihood = -404.72805
Iteration 2:  log likelihood = -404.07242
Iteration 3:  log likelihood = -403.93667
Iteration 4:  log likelihood = -403.90462
Iteration 5:  log likelihood = -403.89822
Iteration 6:  log likelihood = -403.89706
Iteration 7:  log likelihood = -403.89694
Iteration 8:  log likelihood = -403.89692

```

```

Multinomial logistic regression      Number of obs   =      402
                                     LR chi2(18)        =      34.58
                                     Prob > chi2         =      0.0107
Log likelihood = -403.89692          Pseudo R2        =      0.0410

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	Asian	(omitted)					
	Black	4.142282	4.62752	1.27	0.203	.463795	36.99587

Latino		.6843826	.7026336	-0.37	0.712	.0914951	5.119178
Other		1633929	1.40e+09	0.02	0.987	0	.
AsianMale		(omitted)					
BlackMale		.6907376	4294.559	-0.00	1.000	0	.
LatinoMale		1.78741	2.57597	0.40	0.687	.106046	30.12689
OtherMale		2.427575	3281.663	0.00	0.999	0	.
male		.5140659	.1252885	-2.73	0.006	.3188331	.8288467
tenure		.9702988	.0147324	-1.99	0.047	.9418491	.9996078
age		.9940354	.0161192	-0.37	0.712	.9629392	1.026136

2		(omitted)					
Asian		(omitted)					
Black		3.935882	4.932571	1.09	0.274	.3375123	45.89809
Latino		1.74e-06	.0012668	-0.02	0.985	0	.
Other		1.094033	1677.02	0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		1.56e+07	6.68e+10	0.00	0.997	0	.
LatinoMale		1171211	8.53e+08	0.02	0.985	0	.
OtherMale		1.067143	2594.788	0.00	1.000	0	.
male		.8667342	.2638938	-0.47	0.639	.477223	1.574166
tenure		.9999982	.0183308	-0.00	1.000	.9647081	1.036579
age		.9795933	.0198325	-1.02	0.308	.9414835	1.019246

3 | (base outcome)

-----> region = 17

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -410.96404  
Iteration 1: log likelihood = -391.61195  
Iteration 2: log likelihood = -390.51416  
Iteration 3: log likelihood = -390.29486  
Iteration 4: log likelihood = -390.24158  
Iteration 5: log likelihood = -390.23117  
Iteration 6: log likelihood = -390.22933  
Iteration 7: log likelihood = -390.22914  
Iteration 8: log likelihood = -390.2291  
Iteration 9: log likelihood = -390.22909

Multinomial logistic regression	Number of obs	=	414
	LR chi2(18)	=	41.47
	Prob > chi2	=	0.0013
Log likelihood = -390.22909	Pseudo R2	=	0.0505

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1						
Asian		(omitted)				
Black		.2636465	.2295038	-1.53	0.126	.0478688 1.452083
Latino		.8044571	.5480822	-0.32	0.749	.2116281 3.057964
Other		.5013226	.718584	-0.48	0.630	.0302013 8.32165
AsianMale		(omitted)				
BlackMale		1.35e-06	.0018379	-0.01	0.992	0
LatinoMale		.8542028	.7534099	-0.18	0.858	.1516346 4.811979
OtherMale		5.224821	9.882426	0.87	0.382	.1282556 212.8464
male		.3865911	.0936736	-3.92	0.000	.2404363 .6215896
tenure		.9800204	.0141318	-1.40	0.162	.9527104 1.008113
age		.9954624	.0159444	-0.28	0.776	.9646974 1.027208

2		(omitted)				
Asian		(omitted)				
Black		1.008949	1.174771	0.01	0.994	.1029847 9.884761
Latino		7.18e-07	.0007273	-0.01	0.989	0

Other		7.39e-07	.0016475	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		2.63e-07	.0006278	-0.01	0.995	0	.
LatinoMale		2357181	2.39e+09	0.01	0.988	0	.
OtherMale		1.327548	3876.656	0.00	1.000	0	.
male		1.085859	.4149959	0.22	0.829	.5134043	2.296611
tenure		.9927626	.021	-0.34	0.731	.952445	1.034787
age		.9671903	.0225345	-1.43	0.152	.9240166	1.012381
-----							
3		(base outcome)					
-----							

-----> region = 18

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -395.09289  
Iteration 1: log likelihood = -382.84209  
Iteration 2: log likelihood = -381.44942  
Iteration 3: log likelihood = -380.90804  
Iteration 4: log likelihood = -380.87727  
Iteration 5: log likelihood = -380.87421  
Iteration 6: log likelihood = -380.87351  
Iteration 7: log likelihood = -380.87336  
Iteration 8: log likelihood = -380.87333

Multinomial logistic regression	Number of obs	=	407
	LR chi2(14)	=	28.44
	Prob > chi2	=	0.0124
Log likelihood = -380.87333	Pseudo R2	=	0.0360

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
Asian		(omitted)				
Black		6.166182	9.087996	1.23	0.217	.3431446 110.804
Latino		2.655416	2.047757	1.27	0.205	.5857578 12.0378
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		427514.2	2.40e+08	0.02	0.982	0 .
LatinoMale		.773468	.702105	-0.28	0.777	.1305497 4.582568
OtherMale		(omitted)				
male		2.065324	.8751731	1.71	0.087	.9001083 4.738944
tenure		1.041241	.023968	1.76	0.079	.9953085 1.089293
age		.933449	.0274174	-2.34	0.019	.8812293 .9887631
-----						
3						
Asian		(omitted)				
Black		3.337636	4.161978	0.97	0.334	.2897373 38.44798
Latino		2.093356	1.107761	1.40	0.163	.7420015 5.905838
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		207560.9	1.16e+08	0.02	0.983	0 .
LatinoMale		.4579876	.3033755	-1.18	0.238	.1250286 1.677637
OtherMale		(omitted)				
male		1.773215	.4436631	2.29	0.022	1.085896 2.895571
tenure		1.021095	.0138287	1.54	0.123	.9943479 1.048562
age		.9736239	.0168299	-1.55	0.122	.9411904 1.007175
-----						

-----> region = 19

```
Iteration 0: log likelihood = -679.29815
Iteration 1: log likelihood = -642.46167
Iteration 2: log likelihood = -641.62958
Iteration 3: log likelihood = -641.5068
Iteration 4: log likelihood = -641.48505
Iteration 5: log likelihood = -641.48294
Iteration 6: log likelihood = -641.48249
Iteration 7: log likelihood = -641.48239
Iteration 8: log likelihood = -641.48237
```

```
Multinomial logistic regression      Number of obs   =      638
                                     LR chi2(22)        =      75.63
                                     Prob > chi2         =      0.0000
Log likelihood = -641.48237          Pseudo R2        =      0.0557
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	1.28e-06	.0013731	-0.01	0.990	0	.
	Black	.9434117	.5695173	-0.10	0.923	.2889677	3.080018
	Latino	1.315695	.3662578	0.99	0.324	.7624322	2.270436
	Other	2.05e-06	.0021878	-0.01	0.990	0	.
	AsianMale	1.26e+12	1.68e+15	0.02	0.983	0	.
	BlackMale	1.107405	1.070093	0.11	0.916	.1666418	7.359169
	LatinoMale	1.28327	.4979223	0.64	0.520	.5998499	2.745321
	OtherMale	322010.7	9.11e+08	0.00	0.996	0	.
	male	1.205691	.3129253	0.72	0.471	.724962	2.005197
	tenure	1.043733	.0127116	3.51	0.000	1.019114	1.068947
	age	.9721918	.0129112	-2.12	0.034	.9472128	.9978295
3							
	Asian	1.32e-06	.0016667	-0.01	0.991	0	.
	Black	.3448872	.2864341	-1.28	0.200	.0677237	1.756358
	Latino	1.532369	.4743196	1.38	0.168	.8353874	2.810857
	Other	1.90e-06	.0024091	-0.01	0.992	0	.
	AsianMale	751385.7	1.40e+09	0.01	0.994	0	.
	BlackMale	4.208451	4.79074	1.26	0.207	.4520088	39.18299
	LatinoMale	1.03484	.4539416	0.08	0.938	.4380125	2.444895
	OtherMale	1.11e+12	2.35e+15	0.01	0.990	0	.
	male	1.189189	.3534185	0.58	0.560	.6641708	2.129227
	tenure	1.095157	.0158478	6.28	0.000	1.064533	1.126663
	age	.9816698	.0153933	-1.18	0.238	.9519584	1.012308

-----> region = 20

```
Iteration 0: log likelihood = -1876.0007
Iteration 1: log likelihood = -1788.6881
Iteration 2: log likelihood = -1785.8574
Iteration 3: log likelihood = -1785.4628
Iteration 4: log likelihood = -1785.3949
Iteration 5: log likelihood = -1785.3797
Iteration 6: log likelihood = -1785.376
Iteration 7: log likelihood = -1785.3752
Iteration 8: log likelihood = -1785.3751
Iteration 9: log likelihood = -1785.375
Iteration 10: log likelihood = -1785.375
```

```
Multinomial logistic regression      Number of obs   =     1939
                                     LR chi2(22)        =     181.25
                                     Prob > chi2         =     0.0000
```

Log likelihood = -1785.375 Pseudo R2 = 0.0483

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	Asian	5.07e-07	.0005639	-0.01	0.990	0	.
	Black	1.353485	.4098955	1.00	0.318	.7476041	2.45039
	Latino	.8686935	.1588677	-0.77	0.441	.6070128	1.243184
	Other	4.09e-07	.000565	-0.01	0.992	0	.
	AsianMale	.8979386	2529.12	-0.00	1.000	0	.
	BlackMale	1.73026	.7935424	1.20	0.232	.7042518	4.251034
	LatinoMale	1.855904	.4949231	2.32	0.020	1.100427	3.130038
	OtherMale	5006401	6.92e+09	0.01	0.991	0	.
	male	1.538661	.2329411	2.85	0.004	1.14361	2.07018
	tenure	1.009658	.0086244	1.13	0.260	.9928956	1.026704
	age	.9719337	.0083611	-3.31	0.001	.9556836	.9884602
3							
	Asian	1.456071	1.250392	0.44	0.662	.270532	7.836942
	Black	1.032746	.3331592	0.10	0.920	.5487818	1.943512
	Latino	.9636852	.1807697	-0.20	0.844	.6672144	1.39189
	Other	1.371935	1.554294	0.28	0.780	.1489323	12.63799
	AsianMale	1.807262	3.013919	0.35	0.723	.0687855	47.48379
	BlackMale	.7744493	.4137723	-0.48	0.632	.2717775	2.206848
	LatinoMale	1.343646	.3642787	1.09	0.276	.7897938	2.285893
	OtherMale	4.21e-07	.0005337	-0.01	0.991	0	.
	male	1.796452	.2737374	3.84	0.000	1.332639	2.421692
	tenure	1.071262	.0090685	8.13	0.000	1.053635	1.089185
	age	.9929214	.0089002	-0.79	0.428	.9756297	1.01052

### 1993-94

. mlogit admin male age tenure Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale, rrr iter(20)

Iteration 0: log likelihood = -49681.853  
Iteration 1: log likelihood = -47672.742  
Iteration 2: log likelihood = -44782.943  
Iteration 3: log likelihood = -44729.501  
Iteration 4: log likelihood = -44728.791  
Iteration 5: log likelihood = -44728.791

Multinomial logistic regression Number of obs = 235571  
LR chi2(22) = 9906.12  
Prob > chi2 = 0.0000  
Log likelihood = -44728.791 Pseudo R2 = 0.0997

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	3.609767	.1331056	34.81	0.000	3.358089	3.880308
	age	1.02621	.0019514	13.61	0.000	1.022393	1.030042
	tenure	1.035073	.0021275	16.77	0.000	1.030911	1.039251
	Asian	.2529606	.1137137	-3.06	0.002	.1048121	.6105124
	Black	1.764954	.1083533	9.25	0.000	1.564865	1.990627
	Latino	1.572002	.0902386	7.88	0.000	1.404724	1.759199
	Other	.4136209	.1470589	-2.48	0.013	.2060468	.8303076
	AsianMale	2.935123	1.52177	2.08	0.038	1.062441	8.108639
	BlackMale	1.132055	.1010257	1.39	0.165	.9503974	1.348433
	LatinoMale	.957656	.07547	-0.55	0.583	.820595	1.11761



OtherMale		1.796656	.8295576	1.27	0.204	.7268513	4.441037
-----+-----							
3							
male		4.926821	.1563489	50.25	0.000	4.629718	5.242989
age		1.048783	.0018281	27.33	0.000	1.045206	1.052372
tenure		1.054944	.0018329	30.78	0.000	1.051358	1.058543
Asian		.301407	.0966309	-3.74	0.000	.1607898	.5649996
Black		.9865679	.0664855	-0.20	0.841	.8644979	1.125875
Latino		1.520399	.0827874	7.69	0.000	1.366497	1.691634
Other		.2993705	.1137569	-3.17	0.002	.1421554	.6304556
AsianMale		1.376043	.5478087	0.80	0.423	.6306157	3.002611
BlackMale		.7809597	.0779421	-2.48	0.013	.6422086	.9496883
LatinoMale		.5875052	.0446815	-6.99	0.000	.5061454	.6819431
OtherMale		2.703563	1.205101	2.23	0.026	1.128541	6.476728
-----+-----							

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=235571)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----+-----				
male		3490.464	2	0.000
age		864.031	2	0.000
tenure		1196.305	2	0.000
Asian		35.495	2	0.000
Black		76.513	2	0.000
Latino		109.081	2	0.000
Other		23.582	2	0.000
AsianMale		5.541	2	0.063
BlackMale		8.600	2	0.014
LatinoMale		48.540	2	0.000
OtherMale		7.139	2	0.028
-----+-----				

\*\*\*\* Wald tests for independent variables (N=235571)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----+-----				
male		3593.649	2	0.000
age		900.974	2	0.000
tenure		1166.918	2	0.000
Asian		23.107	2	0.000
Black		85.873	2	0.000
Latino		118.388	2	0.000
Other		16.145	2	0.000
AsianMale		4.865	2	0.088
BlackMale		8.581	2	0.014
LatinoMale		48.950	2	0.000
OtherMale		6.482	2	0.039
-----+-----				

\*\*\*\* Hausman tests of IIA assumption (N=235571)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
-----+-----					
2		-81.056	12	---	---
3		158.973	12	0.000	against Ho
-----+-----					

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=235571)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-1.18e+04	-1.18e+04	18.358	12	0.105	for Ho
3	-1.05e+04	-1.05e+04	3.681	12	0.989	for Ho

\*\*\*\* Wald tests for combining alternatives (N=235571)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	464.414	11	0.000
2-	1	3366.448	11	0.000
3-	1	6710.048	11	0.000

\*\*\*\* LR tests for combining alternatives (N=235571)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	472.724	11	0.000
2-	1	3232.772	11	0.000
3-	1	6965.482	11	0.000

by region

. bysort region: mlogit admin male age tenure Asian Black Latino Other AsianMale  
BlackMale LatinoMale OtherMale, rrr iter(20)

-----> region = 1

Iteration 0: log likelihood = -3413.3027  
Iteration 1: log likelihood = -3323.8857  
Iteration 2: log likelihood = -3035.1848  
Iteration 3: log likelihood = -3029.2104  
Iteration 4: log likelihood = -3028.8133  
Iteration 5: log likelihood = -3028.7425  
Iteration 6: log likelihood = -3028.7266  
Iteration 7: log likelihood = -3028.7227  
Iteration 8: log likelihood = -3028.7219  
Iteration 9: log likelihood = -3028.7217  
Iteration 10: log likelihood = -3028.7217  
Iteration 11: log likelihood = -3028.7217

Multinomial logistic regression                      Number of obs        =        16781  
   LR chi2(22)        =        769.16  
   Prob > chi2        =        0.0000  
Log likelihood = -3028.7217                      Pseudo R2        =        0.1127

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

2							
	male	3.158355	.7677962	4.73	0.000	1.961254	5.086137
	age	1.026905	.0072343	3.77	0.000	1.012824	1.041182
	tenure	1.045828	.0081895	5.72	0.000	1.029899	1.062003
	Asian	5.84e-07	.0009563	-0.01	0.993	0	.
	Black	5.63e-07	.0016124	-0.01	0.996	0	.
	Latino	1.609949	.3408485	2.25	0.024	1.063165	2.437941
	Other	5.24e-07	.001304	-0.01	0.995	0	.
	AsianMale	.326052	1170.106	-0.00	1.000	0	.
	BlackMale	2668378	7.64e+09	0.01	0.996	0	.
	LatinoMale	1.214978	.3314985	0.71	0.475	.711744	2.07402
	OtherMale	2090715	5.21e+09	0.01	0.995	0	.
3							
	male	.9884276	.2651874	-0.04	0.965	.584217	1.672305
	age	1.036657	.0081158	4.60	0.000	1.020872	1.052687
	tenure	1.092495	.0086065	11.23	0.000	1.075756	1.109494
	Asian	7.84e-07	.0009744	-0.01	0.991	0	.
	Black	7.09e-07	.0015481	-0.01	0.995	0	.
	Latino	1.06237	.1877835	0.34	0.732	.7513082	1.502219
	Other	5.52e-07	.0011402	-0.01	0.994	0	.
	AsianMale	.8995535	2774.242	-0.00	1.000	0	.
	BlackMale	.9462079	2915.231	-0.00	1.000	0	.
	LatinoMale	2.934934	.8695455	3.63	0.000	1.642136	5.245507
	OtherMale	1.26e+07	2.60e+10	0.01	0.994	0	.

-----> region = 2

Iteration 0: log likelihood = -1708.796  
Iteration 1: log likelihood = -1642.9812  
Iteration 2: log likelihood = -1534.5372  
Iteration 3: log likelihood = -1533.1376  
Iteration 4: log likelihood = -1533.0587  
Iteration 5: log likelihood = -1533.0422  
Iteration 6: log likelihood = -1533.0387  
Iteration 7: log likelihood = -1533.0379  
Iteration 8: log likelihood = -1533.0377  
Iteration 9: log likelihood = -1533.0377  
Iteration 10: log likelihood = -1533.0377

Multinomial logistic regression	Number of obs	=	7562
	LR chi2(22)	=	351.52
	Prob > chi2	=	0.0000
Log likelihood = -1533.0377	Pseudo R2	=	0.1029

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	3.1572	.7420166	4.89	0.000	1.991818	5.004429
	age	1.00257	.0118352	0.22	0.828	.9796398	1.026037
	tenure	1.05913	.0127164	4.78	0.000	1.034497	1.084349
	Asian	2.30e-06	.0060324	-0.00	0.996	0	.
	Black	3.355884	1.812703	2.24	0.025	1.164193	9.67362
	Latino	1.077921	.2745899	0.29	0.768	.6542614	1.775915
	Other	4.510903	3.371412	2.02	0.044	1.042515	19.51842
	AsianMale	.2312908	1003.876	-0.00	1.000	0	.
	BlackMale	.7266306	.6048603	-0.38	0.701	.1421537	3.714233
	LatinoMale	1.343191	.4710134	0.84	0.400	.675537	2.670708
	OtherMale	.5916615	.7662892	-0.41	0.685	.0467349	7.490405
3							
	male	3.345195	.6756841	5.98	0.000	2.251595	4.969957

age		1.043133	.0099965	4.41	0.000	1.023723	1.06291
tenure		1.052504	.0098293	5.48	0.000	1.033414	1.071946
Asian		2.80e-06	.0058119	-0.01	0.995	0	.
Black		.6547913	.6676344	-0.42	0.678	.0887576	4.830588
Latino		1.483893	.311217	1.88	0.060	.9837393	2.238335
Other		1.98e-06	.0017718	-0.01	0.988	0	.
AsianMale		.1801658	578.9417	-0.00	1.000	0	.
BlackMale		3.633442	4.233791	1.11	0.268	.3702318	35.65847
LatinoMale		1.229433	.353602	0.72	0.473	.6996607	2.160341
OtherMale		.3119852	626.4301	-0.00	1.000	0	.

-----> region = 3

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -834.05907  
Iteration 1: log likelihood = -813.00301  
Iteration 2: log likelihood = -723.31421  
Iteration 3: log likelihood = -720.74667  
Iteration 4: log likelihood = -720.33939  
Iteration 5: log likelihood = -720.26759  
Iteration 6: log likelihood = -720.24986  
Iteration 7: log likelihood = -720.24597  
Iteration 8: log likelihood = -720.24503  
Iteration 9: log likelihood = -720.24484  
Iteration 10: log likelihood = -720.24481

Multinomial logistic regression

Number of obs = 4025  
LR chi2(20) = 227.63  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.1365

Log likelihood = -720.24481

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	male	4.400319	1.437514	4.54	0.000	2.319584 8.347534
	age	1.042522	.0180984	2.40	0.016	1.007647 1.078605
	tenure	.978042	.0186917	-1.16	0.245	.9420846 1.015372
	Asian	9.81e-07	.0050925	-0.00	0.998	0 .
	Black	4.459956	2.549574	2.62	0.009	1.454567 13.67501
	Latino	2.269943	1.433381	1.30	0.194	.6584368 7.825564
	Other	6.83e-07	.0022092	-0.00	0.996	0 .
	AsianMale	(omitted)				
	BlackMale	.2426154	.286995	-1.20	0.231	.0238789 2.465035
	LatinoMale	1.270422	1.00501	0.30	0.762	.269511 5.988519
	OtherMale	.0253586	529.7691	-0.00	1.000	0 .
3						
	male	6.867493	1.45506	9.09	0.000	4.533657 10.40274
	age	1.078583	.0132261	6.17	0.000	1.052969 1.10482
	tenure	.9966584	.0114233	-0.29	0.770	.9745187 1.019301
	Asian	1.59e-06	.0050377	-0.00	0.997	0 .
	Black	1.265731	.7794538	0.38	0.702	.3785826 4.231774
	Latino	1.23e-06	.0007025	-0.02	0.981	0 .
	Other	8.30e-07	.0016078	-0.01	0.994	0 .
	AsianMale	(omitted)				
	BlackMale	1.18e-07	.0001669	-0.01	0.991	0 .
	LatinoMale	1542029	8.81e+08	0.02	0.980	0 .
	OtherMale	3244418	6.29e+09	0.01	0.994	0 .

-----> region = 4

```

Iteration 0: log likelihood = -9860.9827
Iteration 1: log likelihood = -9588.5981
Iteration 2: log likelihood = -8922.469
Iteration 3: log likelihood = -8909.2714
Iteration 4: log likelihood = -8909.2089
Iteration 5: log likelihood = -8909.2089

```

Multinomial logistic regression

```

Number of obs   =    46513
LR chi2(22)     =    1903.55
Prob > chi2     =     0.0000
Pseudo R2      =     0.0965

```

Log likelihood = -8909.2089

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	2.874119	.2122375	14.30	0.000	2.486843	3.321706
	age	1.015742	.004014	3.95	0.000	1.007905	1.02364
	tenure	1.051918	.0045717	11.65	0.000	1.042995	1.060916
	Asian	.4005783	.2854718	-1.28	0.199	.0991013	1.619181
	Black	.986595	.0982872	-0.14	0.892	.8115961	1.199328
	Latino	1.388615	.2093723	2.18	0.029	1.033331	1.866053
	Other	.5375286	.3131765	-1.07	0.287	.1715821	1.683958
	AsianMale	.6473701	.8007964	-0.35	0.725	.0573082	7.312879
	BlackMale	1.247871	.1912381	1.44	0.148	.9241063	1.685069
	LatinoMale	1.204206	.2727777	0.82	0.412	.7724778	1.877222
	OtherMale	1.896834	1.582612	0.77	0.443	.3696873	9.732498
3							
	male	3.110095	.2545114	13.87	0.000	2.64921	3.65116
	age	1.026087	.0049666	5.32	0.000	1.016398	1.035867
	tenure	1.105131	.0052944	20.87	0.000	1.094802	1.115556
	Asian	.4280468	.30734	-1.18	0.237	.104791	1.748472
	Black	.508644	.064978	-5.29	0.000	.3959816	.6533605
	Latino	2.740806	.3751337	7.37	0.000	2.095919	3.584117
	Other	.2464985	.2476791	-1.39	0.163	.0343982	1.766415
	AsianMale	1.945762	1.833793	0.71	0.480	.3068111	12.33981
	BlackMale	1.277755	.2458292	1.27	0.203	.8763623	1.862995
	LatinoMale	.5739464	.142117	-2.24	0.025	.3532663	.9324821
	OtherMale	1.758634	2.528576	0.39	0.695	.1050295	29.44688

-----> region = 5

```

Iteration 0: log likelihood = -1244.8612
Iteration 1: log likelihood = -1198.5081
Iteration 2: log likelihood = -1041.1279
Iteration 3: log likelihood = -1037.478
Iteration 4: log likelihood = -1037.2071
Iteration 5: log likelihood = -1037.1622
Iteration 6: log likelihood = -1037.1516
Iteration 7: log likelihood = -1037.1492
Iteration 8: log likelihood = -1037.1486
Iteration 9: log likelihood = -1037.1486
Iteration 10: log likelihood = -1037.1485

```

Multinomial logistic regression

```

Number of obs   =     5841
LR chi2(22)     =     415.43
Prob > chi2     =     0.0000
Pseudo R2      =     0.1669

```

Log likelihood = -1037.1485

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
--	-------	-----	-----------	---	------	----------------------	--

1		(base outcome)					
2	male	7.922359	1.878858	8.73	0.000	4.977193	12.61027
	age	.9958976	.0139458	-0.29	0.769	.968936	1.023609
	tenure	1.073682	.0145675	5.24	0.000	1.045506	1.102616
	Asian	4.2781	4.460471	1.39	0.163	.5543236	33.01706
	Black	1.541879	.5806467	1.15	0.250	.7370578	3.225515
	Latino	5.03515	3.794373	2.15	0.032	1.149652	22.05252
	Other	6.60e-07	.001367	-0.01	0.995	0	.
	AsianMale	2.69e-08	.0001136	-0.00	0.997	0	.
	BlackMale	.6912498	.3383014	-0.75	0.451	.264882	1.803921
	LatinoMale	.2020721	.260953	-1.24	0.216	.0160797	2.539428
	OtherMale	.2030346	1020.004	-0.00	1.000	0	.
3	male	10.04106	2.129749	10.88	0.000	6.625772	15.21679
	age	1.036703	.0123333	3.03	0.002	1.01281	1.06116
	tenure	1.065097	.0117281	5.73	0.000	1.042357	1.088334
	Asian	1.05e-06	.0018914	-0.01	0.994	0	.
	Black	1.614986	.5452845	1.42	0.156	.8332372	3.130176
	Latino	1.05e-06	.0015342	-0.01	0.992	0	.
	Other	6.84e-07	.0011988	-0.01	0.994	0	.
	AsianMale	.1114026	443.2233	-0.00	1.000	0	.
	BlackMale	.418433	.1889022	-1.93	0.054	.1727228	1.013683
	LatinoMale	.0810154	265.6027	-0.00	0.999	0	.
	OtherMale	.1872864	802.5774	-0.00	1.000	0	.

-----> region = 6

Iteration 0: log likelihood = -1744.3763  
Iteration 1: log likelihood = -1587.7512  
Iteration 2: log likelihood = -1543.9359  
Iteration 3: log likelihood = -1541.6345  
Iteration 4: log likelihood = -1541.2431  
Iteration 5: log likelihood = -1541.1913  
Iteration 6: log likelihood = -1541.1844  
Iteration 7: log likelihood = -1541.1832  
Iteration 8: log likelihood = -1541.1829  
Iteration 9: log likelihood = -1541.1828  
Iteration 10: log likelihood = -1541.1828

Multinomial logistic regression	Number of obs	=	8004
	LR chi2(22)	=	406.39
	Prob > chi2	=	0.0000
Log likelihood = -1541.1828	Pseudo R2	=	0.1165

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	4.271775	.7387194	8.40	0.000	3.043755	5.995245
	age	1.035014	.0099942	3.56	0.000	1.01561	1.054789
	tenure	1.024458	.0110887	2.23	0.026	1.002954	1.046424
	Asian	3.68e-07	.0008849	-0.01	0.995	0	.
	Black	.6449916	.3848792	-0.73	0.462	.2002744	2.077221
	Latino	.7030281	.7116585	-0.35	0.728	.0966767	5.112385
	Other	4.98e-07	.0007531	-0.01	0.992	0	.
	AsianMale	.2914901	1341.019	-0.00	1.000	0	.
	BlackMale	4.572489	3.141357	2.21	0.027	1.189506	17.57676
	LatinoMale	3.493855	4.022186	1.09	0.277	.3659206	33.35975
	OtherMale	.2824512	1466.837	-0.00	1.000	0	.

3							
	male	6.091312	.9169283	12.00	0.000	4.535018	8.181683
	age	1.064296	.0089782	7.39	0.000	1.046844	1.08204
	tenure	1.025425	.0091049	2.83	0.005	1.007734	1.043427
	Asian	3.70e-07	.0007445	-0.01	0.994	0	.
	Black	.8316677	.3927106	-0.39	0.696	.3296214	2.09838
	Latino	9.11e-07	.0007745	-0.02	0.987	0	.
	Other	5.55e-07	.0007091	-0.01	0.991	0	.
	AsianMale	.2194318	841.1979	-0.00	1.000	0	.
	BlackMale	1.087084	.690015	0.13	0.895	.3133085	3.771848
	LatinoMale	407927.2	3.47e+08	0.02	0.988	0	.
	OtherMale	3963978	5.07e+09	0.01	0.991	0	.

-----> region = 7

Iteration 0: log likelihood = -2180.4839  
Iteration 1: log likelihood = -2003.7146  
Iteration 2: log likelihood = -1858.9683  
Iteration 3: log likelihood = -1853.2835  
Iteration 4: log likelihood = -1853.0282  
Iteration 5: log likelihood = -1852.9872  
Iteration 6: log likelihood = -1852.9772  
Iteration 7: log likelihood = -1852.9751  
Iteration 8: log likelihood = -1852.9746  
Iteration 9: log likelihood = -1852.9745  
Iteration 10: log likelihood = -1852.9745

Multinomial logistic regression                      Number of obs    =        10726  
   LR chi2(22)        =        655.02  
   Prob > chi2        =        0.0000  
Log likelihood = -1852.9745                          Pseudo R2        =        0.1502

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	male	9.854562	2.249143	10.02	0.000	6.300347 15.41382
	age	1.017252	.0116405	1.49	0.135	.9946914 1.040325
	tenure	1.054917	.0124454	4.53	0.000	1.030804 1.079594
	Asian	5.50314	5.748536	1.63	0.103	.710317 42.63526
	Black	1.35597	.6217979	0.66	0.507	.5519764 3.331039
	Latino	4.005344	4.11198	1.35	0.176	.5355208 29.95734
	Other	4.11e-06	.0038567	-0.01	0.989	0 .
	AsianMale	.4624263	.6975406	-0.51	0.609	.0240469 8.892539
	BlackMale	2.783927	1.44006	1.98	0.048	1.01007 7.672986
	LatinoMale	1.32e-07	.0002084	-0.01	0.992	0 .
	OtherMale	470116.3	4.41e+08	0.01	0.989	0 .
3						
	male	6.33925	.8055224	14.53	0.000	4.941701 8.132035
	age	1.054925	.0073209	7.70	0.000	1.040673 1.069372
	tenure	1.035993	.0075064	4.88	0.000	1.021385 1.05081
	Asian	1.47e-06	.001303	-0.02	0.988	0 .
	Black	.8555714	.2435206	-0.55	0.584	.4897557 1.494628
	Latino	1.010697	1.024442	0.01	0.992	.1386255 7.368834
	Other	2.45e-06	.0014633	-0.02	0.983	0 .
	AsianMale	496805	4.41e+08	0.01	0.988	0 .
	BlackMale	1.301065	.4956162	0.69	0.490	.6166656 2.745037
	LatinoMale	1.111348	1.401519	0.08	0.933	.0938428 13.16131
	OtherMale	278927.3	1.67e+08	0.02	0.983	0 .

-----> region = 8

```
Iteration 0: log likelihood = -818.62377
Iteration 1: log likelihood = -713.80235
Iteration 2: log likelihood = -673.55472
Iteration 3: log likelihood = -672.51715
Iteration 4: log likelihood = -672.39238
Iteration 5: log likelihood = -672.37612
Iteration 6: log likelihood = -672.37392
Iteration 7: log likelihood = -672.37352
Iteration 8: log likelihood = -672.37343
Iteration 9: log likelihood = -672.37341
```

```
Multinomial logistic regression      Number of obs   =      3969
                                     LR chi2(22)      =      292.50
                                     Prob > chi2      =      0.0000
Log likelihood = -672.37341          Pseudo R2       =      0.1787
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	16.41121	6.875658	6.68	0.000	7.219722	37.30447
	age	1.051522	.0193613	2.73	0.006	1.014252	1.090163
	tenure	1.010393	.0189152	0.55	0.581	.9739918	1.048155
	Asian	6.06e-06	.0467775	-0.00	0.999	0	.
	Black	3.648648	2.558392	1.85	0.065	.9231641	14.42066
	Latino	6.95e-06	.0133968	-0.01	0.995	0	.
	Other	5.89e-06	.0077382	-0.01	0.993	0	.
	AsianMale	.0437797	368.5182	-0.00	1.000	0	.
	BlackMale	.7753749	.6405026	-0.31	0.758	.1535932	3.914275
	LatinoMale	.0427542	153.3725	-0.00	0.999	0	.
	OtherMale	.085964	247.2712	-0.00	0.999	0	.
3							
	male	11.15243	2.521667	10.67	0.000	7.159864	17.37137
	age	1.031916	.0129533	2.50	0.012	1.006838	1.057619
	tenure	1.040662	.0131583	3.15	0.002	1.015189	1.066774
	Asian	4.11e-06	.0201305	-0.00	0.998	0	.
	Black	.8239675	.5090238	-0.31	0.754	.2455059	2.765402
	Latino	5.24e-06	.0064897	-0.01	0.992	0	.
	Other	3.99e-06	.0033181	-0.01	0.988	0	.
	AsianMale	.0717415	382.3737	-0.00	1.000	0	.
	BlackMale	1.316676	.9709609	0.37	0.709	.3102947	5.587061
	LatinoMale	.0738926	170.7754	-0.00	0.999	0	.
	OtherMale	300462.2	2.50e+08	0.02	0.988	0	.

-----> region = 9

```
Iteration 0: log likelihood = -572.18827
Iteration 1: log likelihood = -494.9095
Iteration 2: log likelihood = -464.38849
Iteration 3: log likelihood = -463.48238
Iteration 4: log likelihood = -463.37027
Iteration 5: log likelihood = -463.35301
Iteration 6: log likelihood = -463.35099
Iteration 7: log likelihood = -463.35059
Iteration 8: log likelihood = -463.3505
Iteration 9: log likelihood = -463.35048
```

```
Multinomial logistic regression      Number of obs   =      3037
                                     LR chi2(22)      =      217.68
```



Log likelihood = -463.35048

Prob > chi2 = 0.0000  
Pseudo R2 = 0.1902

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	11.03259	5.673143	4.67	0.000	4.026942	30.22592
	age	1.062604	.0319505	2.02	0.043	1.001792	1.127108
	tenure	1.055139	.0288425	1.96	0.050	1.000096	1.113211
	Asian	4.44e-06	.0152109	-0.00	0.997	0	.
	Black	8.012254	9.039488	1.84	0.065	.8778544	73.12854
	Latino	5.66e-06	.0090386	-0.01	0.994	0	.
	Other	5.66e-06	.0117671	-0.01	0.995	0	.
	AsianMale	.3215295	3319.218	-0.00	1.000	0	.
	BlackMale	4.19e-08	.0001055	-0.01	0.995	0	.
	LatinoMale	1319582	2.11e+09	0.01	0.993	0	.
	OtherMale	.0872925	304.0951	-0.00	0.999	0	.
3							
	male	9.753389	2.379129	9.34	0.000	6.046767	15.73214
	age	1.070128	.0154186	4.70	0.000	1.040331	1.100778
	tenure	1.022973	.01408	1.65	0.099	.9957452	1.050945
	Asian	4.43e-06	.0079011	-0.01	0.994	0	.
	Black	5.811441	3.812552	2.68	0.007	1.606421	21.02365
	Latino	3.358177	3.559228	1.14	0.253	.4206697	26.80809
	Other	4.05e-06	.00417	-0.01	0.990	0	.
	AsianMale	.2366882	1216.257	-0.00	1.000	0	.
	BlackMale	5.86e-08	.0000076	-0.01	0.990	0	.
	LatinoMale	3.76e-08	.0001341	-0.00	0.996	0	.
	OtherMale	.0987912	170.2017	-0.00	0.999	0	.

-----> region = 10

Iteration 0: log likelihood = -6545.4666  
Iteration 1: log likelihood = -6291.0691  
Iteration 2: log likelihood = -5775.8871  
Iteration 3: log likelihood = -5765.6865  
Iteration 4: log likelihood = -5765.575  
Iteration 5: log likelihood = -5765.5663  
Iteration 6: log likelihood = -5765.5646  
Iteration 7: log likelihood = -5765.5643  
Iteration 8: log likelihood = -5765.5642  
Iteration 9: log likelihood = -5765.5642

Multinomial logistic regression

Number of obs = 30222  
LR chi2(22) = 1559.80  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.1192

Log likelihood = -5765.5642

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	4.410659	.4322095	15.14	0.000	3.639926	5.344591
	age	1.016107	.0054216	2.99	0.003	1.005536	1.026789
	tenure	1.060596	.00597	10.45	0.000	1.048959	1.072361
	Asian	.0755832	.0760214	-2.57	0.010	.0105266	.5427023
	Black	2.278732	.3101744	6.05	0.000	1.745139	2.975475
	Latino	3.163143	.6287885	5.79	0.000	2.142456	4.670096
	Other	.3241272	.3257639	-1.12	0.262	.0452081	2.323888

AsianMale		6.844531	7.285721	1.81	0.071	.849718	55.13313
BlackMale		.751193	.1554495	-1.38	0.167	.5007334	1.126929
LatinoMale		.7914484	.2307096	-0.80	0.422	.4469869	1.401362
OtherMale		6.47e-06	.00293	-0.03	0.979	0	.
-----							
3							
male		4.265093	.3770795	16.41	0.000	3.586519	5.072052
age		1.031804	.0053278	6.06	0.000	1.021414	1.042299
tenure		1.085527	.0056089	15.88	0.000	1.074589	1.096576
Asian		.2211392	.093428	-3.57	0.000	.0966153	.5061575
Black		1.278333	.185732	1.69	0.091	.9615477	1.699484
Latino		2.516387	.5134211	4.52	0.000	1.686964	3.753608
Other		.246871	.248271	-1.39	0.164	.0343907	1.772143
AsianMale		2.026319	1.02803	1.39	0.164	.7496527	5.47716
BlackMale		.716493	.1647344	-1.45	0.147	.4565682	1.124393
LatinoMale		.7175964	.221898	-1.07	0.283	.3914428	1.315504
OtherMale		9.462505	10.4021	2.04	0.041	1.097193	81.6073
-----							

-----> region = 11

Iteration 0: log likelihood = -4731.7788  
Iteration 1: log likelihood = -4564.7988  
Iteration 2: log likelihood = -4227.3397  
Iteration 3: log likelihood = -4220.9675  
Iteration 4: log likelihood = -4220.8272  
Iteration 5: log likelihood = -4220.7971  
Iteration 6: log likelihood = -4220.7897  
Iteration 7: log likelihood = -4220.7881  
Iteration 8: log likelihood = -4220.7878  
Iteration 9: log likelihood = -4220.7877  
Iteration 10: log likelihood = -4220.7877

Multinomial logistic regression	Number of obs	=	21086
	LR chi2(22)	=	1021.98
	Prob > chi2	=	0.0000
Log likelihood = -4220.7877	Pseudo R2	=	0.1080

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1			(base outcome)				
2							
	male		3.13017	.3376119	10.58	0.000	2.533726 3.867019
	age		1.069337	.0053892	13.30	0.000	1.058826 1.079952
	tenure		.9136006	.0073811	-11.18	0.000	.8992479 .9281825
	Asian		2.58e-06	.0018014	-0.02	0.985	0 .
	Black		3.14054	.5776677	6.22	0.000	2.189955 4.503742
	Latino		3.110969	.692867	5.10	0.000	2.010566 4.813635
	Other		.3179191	.3199184	-1.14	0.255	.0442346 2.284921
	AsianMale		317409.7	2.22e+08	0.02	0.986	0 .
	BlackMale		.8149341	.2282524	-0.73	0.465	.4706652 1.411019
	LatinoMale		.7329103	.2693374	-0.85	0.398	.3566503 1.506118
	OtherMale		1.71e-06	.0014394	-0.02	0.987	0 .
3							
	male		4.757313	.4641235	15.99	0.000	3.929329 5.759767
	age		1.097464	.0056909	17.94	0.000	1.086367 1.108675
	tenure		.9676504	.0057784	-5.51	0.000	.9563909 .9790424
	Asian		1.302416	1.333355	0.26	0.796	1.1751169 9.686605
	Black		1.839754	.3925903	2.86	0.004	1.210931 2.795117
	Latino		1.727932	.5245149	1.80	0.072	.9531114 3.132632
	Other		.3030549	.305122	-1.19	0.236	.0421225 2.180363
	AsianMale		3.18e-07	.0003839	-0.01	0.990	0 .
	BlackMale		.3768217	.134775	-2.73	0.006	.1869361 .7595891

LatinoMale		.7755796	.3483155	-0.57	0.571	.3216243	1.870269
OtherMale		1.819718	2.275311	0.48	0.632	.156925	21.10164

-----> region = 12

```

Iteration 0: log likelihood = -1775.1807
Iteration 1: log likelihood = -1598.6825
Iteration 2: log likelihood = -1547.8287
Iteration 3: log likelihood = -1545.7472
Iteration 4: log likelihood = -1545.4066
Iteration 5: log likelihood = -1545.3567
Iteration 6: log likelihood = -1545.3506
Iteration 7: log likelihood = -1545.3495
Iteration 8: log likelihood = -1545.3492
Iteration 9: log likelihood = -1545.3492
Iteration 10: log likelihood = -1545.3491

```

Multinomial logistic regression		Number of obs	=	8490
		LR chi2(22)	=	459.66
		Prob > chi2	=	0.0000
Log likelihood = -1545.3491		Pseudo R2	=	0.1295

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	male	4.478516	.9348385	7.18	0.000	2.974785 6.742371
	age	1.009626	.0116827	0.83	0.408	.9869858 1.032785
	tenure	1.050747	.0135382	3.84	0.000	1.024545 1.07762
	Asian	3.94e-07	.000793	-0.01	0.994	0 .
	Black	2.45624	.8824267	2.50	0.012	1.214699 4.966755
	Latino	.9023354	.9164171	-0.10	0.919	.1232772 6.604702
	Other	7.15e-07	.0012646	-0.01	0.994	0 .
	AsianMale	5543009	1.11e+10	0.01	0.994	0 .
	BlackMale	1.305077	.655593	0.53	0.596	.4875772 3.493246
	LatinoMale	3.982437	4.495545	1.22	0.221	.435789 36.39331
	OtherMale	2978956	5.27e+09	0.01	0.993	0 .
3						
	male	6.416823	.9280974	12.85	0.000	4.832885 8.519883
	age	1.050653	.008555	6.07	0.000	1.034019 1.067556
	tenure	1.049085	.0088866	5.66	0.000	1.031811 1.066648
	Asian	3.32e-07	.0004411	-0.01	0.991	0 .
	Black	1.365039	.4376728	0.97	0.332	.7281558 2.558975
	Latino	3.147436	1.376104	2.62	0.009	1.335978 7.415055
	Other	7.93e-07	.000959	-0.01	0.991	0 .
	AsianMale	.1852318	500.4684	-0.00	1.000	0 .
	BlackMale	.7983081	.3816673	-0.47	0.638	.3127628 2.037633
	LatinoMale	4.63e-08	.0000659	-0.01	0.991	0 .
	OtherMale	.1968845	510.7981	-0.00	1.000	0 .

-----> region = 13

```

Iteration 0: log likelihood = -3076.027
Iteration 1: log likelihood = -3033.9505
Iteration 2: log likelihood = -2837.251
Iteration 3: log likelihood = -2796.389
Iteration 4: log likelihood = -2790.4031
Iteration 5: log likelihood = -2789.1758
Iteration 6: log likelihood = -2789.1166
Iteration 7: log likelihood = -2789.113

```

Iteration 8: log likelihood = -2789.1122  
 Iteration 9: log likelihood = -2789.112  
 Iteration 10: log likelihood = -2789.112

Multinomial logistic regression      Number of obs    =      14853  
    LR chi2(22)       =      573.83  
    Prob > chi2       =      0.0000  
 Log likelihood = -2789.112            Pseudo R2        =      0.0933

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	male	3.039356	.4461781	7.57	0.000	2.279424	4.052641
	age	1.038612	.0081055	4.85	0.000	1.022846	1.05462
	tenure	1.034193	.009149	3.80	0.000	1.016416	1.052281
	Asian	2.10e-06	.0021285	-0.01	0.990	0	.
	Black	2.916812	.7346295	4.25	0.000	1.780428	4.778509
	Latino	2.468131	.5407233	4.12	0.000	1.606516	3.79185
	Other	2.14e-06	.0013037	-0.02	0.983	0	.
	AsianMale	.252701	568.4224	-0.00	1.000	0	.
	BlackMale	1.535885	.5968268	1.10	0.269	.7171201	3.289468
	LatinoMale	.7139426	.2543692	-0.95	0.344	.3551324	1.435279
	OtherMale	518525.9	3.16e+08	0.02	0.983	0	.
3							
	male	4.457917	.5381505	12.38	0.000	3.518655	5.647904
	age	1.061159	.0071078	8.86	0.000	1.047319	1.075182
	tenure	1.031094	.0076024	4.15	0.000	1.016301	1.046103
	Asian	2.24e-06	.0018473	-0.02	0.987	0	.
	Black	1.727916	.4852408	1.95	0.051	.996517	2.996131
	Latino	1.77153	.415066	2.44	0.015	1.119215	2.804035
	Other	.5789458	.5869592	-0.54	0.590	.0793697	4.223002
	AsianMale	.1562218	291.9318	-0.00	0.999	0	.
	BlackMale	1.329441	.543175	0.70	0.486	.5968847	2.961065
	LatinoMale	.6368829	.2230791	-1.29	0.198	.3205618	1.26534
	OtherMale	1.22e-06	.0011315	-0.01	0.988	0	.

-----> region = 14

Iteration 0: log likelihood = -742.60795  
 Iteration 1: log likelihood = -726.5935  
 Iteration 2: log likelihood = -630.27298  
 Iteration 3: log likelihood = -627.73635  
 Iteration 4: log likelihood = -627.439  
 Iteration 5: log likelihood = -627.39044  
 Iteration 6: log likelihood = -627.37824  
 Iteration 7: log likelihood = -627.37573  
 Iteration 8: log likelihood = -627.37519  
 Iteration 9: log likelihood = -627.37507  
 Iteration 10: log likelihood = -627.37504  
 Iteration 11: log likelihood = -627.37504

Multinomial logistic regression      Number of obs    =      3898  
    LR chi2(22)       =      230.47  
    Prob > chi2       =      0.0000  
 Log likelihood = -627.37504            Pseudo R2        =      0.1552

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						

```

2      |
  male | 6.09642 2.284995 4.82 0.000 2.924399 12.70905
  age  | 1.00149 .0227175 0.07 0.948 .9579399 1.04702
  tenure | 1.053507 .0256017 2.14 0.032 1.004505 1.1049
  Asian | 5.55e-07 .0051766 -0.00 0.999 0
  Black | 4.688427 4.987283 1.45 0.146 .582864 37.71266
  Latino | 5.31e-07 .0013096 -0.01 0.995 0
  Other | 4.71e-07 .00199 -0.00 0.997 0
  AsianMale | .1774332 2405.394 -0.00 1.000 0
  BlackMale | .4163758 .6240618 -0.58 0.559 .0220653 7.85709
  LatinoMale | 5334149 1.32e+10 0.01 0.995 0
  OtherMale | .0423681 741.896 -0.00 1.000 0
-----

```

```

3      |
  male | 9.113677 1.989395 10.12 0.000 5.941384 13.97976
  age  | 1.057401 .0129769 4.55 0.000 1.03227 1.083143
  tenure | 1.02991 .0126862 2.39 0.017 1.005344 1.055078
  Asian | 1.05e-06 .0049112 -0.00 0.998 0
  Black | 7.72e-07 .0011971 -0.01 0.993 0
  Latino | 8.30e-07 .0010741 -0.01 0.991 0
  Other | 2.412189 2.518506 0.84 0.399 .3116685 18.66937
  AsianMale | .0778681 547.8446 -0.00 1.000 0
  BlackMale | 1390125 2.16e+09 0.01 0.993 0
  LatinoMale | 971274.9 1.26e+09 0.01 0.992 0
  OtherMale | .6407733 .999347 -0.29 0.775 .0301422 13.62179
-----

```

-----> region = 15

```

Iteration 0: log likelihood = -767.73012
Iteration 1: log likelihood = -735.66874
Iteration 2: log likelihood = -660.64668
Iteration 3: log likelihood = -659.03553
Iteration 4: log likelihood = -658.97078
Iteration 5: log likelihood = -658.95627
Iteration 6: log likelihood = -658.95281
Iteration 7: log likelihood = -658.9521
Iteration 8: log likelihood = -658.95198
Iteration 9: log likelihood = -658.95196

```

```

Multinomial logistic regression      Number of obs   =      3742
                                     LR chi2(22)        =      217.56
                                     Prob > chi2         =      0.0000
Log likelihood = -658.95196          Pseudo R2        =      0.1417

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          |      (base outcome)
-----+-----
2          |
  male | 3.383308 1.344172 3.07 0.002 1.55297 7.370889
  age  | 1.008881 .0238106 0.37 0.708 .9632759 1.056645
  tenure | 1.073471 .0249713 3.05 0.002 1.025627 1.123547
  Asian | 1.77e-06 .0081616 -0.00 0.998 0
  Black | 3.19e-06 .004978 -0.01 0.994 0
  Latino | 2.728436 1.486849 1.84 0.065 .9376814 7.939116
  Other | 2.71e-06 .0047987 -0.01 0.994 0
  AsianMale | .7099798 7129.797 -0.00 1.000 0
  BlackMale | .2410717 883.2143 -0.00 1.000 0
  LatinoMale | .7204798 .5011983 -0.47 0.637 .184284 2.816801
  OtherMale | .1499115 831.2928 -0.00 1.000 0
-----
3          |
  male | 8.186861 1.884938 9.13 0.000 5.213596 12.85575
  age  | 1.070117 .0126707 5.72 0.000 1.045569 1.095242

```

tenure		1.01645	.0120426	1.38	0.168	.9931186	1.040329
Asian		2.45e-06	.0063964	-0.00	0.996	0	.
Black		3.146928	3.320593	1.09	0.277	.3978425	24.89215
Latino		1.547905	.7660263	0.88	0.377	.5868161	4.083066
Other		3.15e-06	.0033526	-0.01	0.991	0	.
AsianMale		.3763343	2138.67	-0.00	1.000	0	.
BlackMale		1.65e-07	.0002867	-0.01	0.993	0	.
LatinoMale		.5244988	.3032987	-1.12	0.264	.1688594	1.629159
OtherMale		505858.3	5.39e+08	0.01	0.990	0	.

-----> region = 16

```

Iteration 0: log likelihood = -1214.5715
Iteration 1: log likelihood = -1194.8345
Iteration 2: log likelihood = -1067.3387
Iteration 3: log likelihood = -1064.6218
Iteration 4: log likelihood = -1064.4877
Iteration 5: log likelihood = -1064.4553
Iteration 6: log likelihood = -1064.4488
Iteration 7: log likelihood = -1064.4475
Iteration 8: log likelihood = -1064.4471
Iteration 9: log likelihood = -1064.4471
Iteration 10: log likelihood = -1064.447

```

Multinomial logistic regression	Number of obs	=	5734
	LR chi2(22)	=	300.25
	Prob > chi2	=	0.0000
Log likelihood = -1064.447	Pseudo R2	=	0.1236

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		5.370784	1.319113	6.84	0.000	3.318753 8.691613
age		1.021056	.0153053	1.39	0.164	.9914946 1.051499
tenure		1.031626	.0167045	1.92	0.054	.9994002 1.064892
Asian		4.98e-07	.0024866	-0.00	0.998	0 .
Black		11.03887	7.069838	3.75	0.000	3.146103 38.73256
Latino		1.305914	1.336387	0.26	0.794	.175732 9.704616
Other		5.53e-07	.0012443	-0.01	0.995	0 .
AsianMale		.2504978	3175.82	-0.00	1.000	0 .
BlackMale		.1754553	.2172713	-1.41	0.160	.0154917 1.987164
LatinoMale		1.991483	2.385954	0.57	0.565	.1902699 20.8441
OtherMale		.183667	1041.751	-0.00	1.000	0 .
3						
male		7.561831	1.257194	12.17	0.000	5.458964 10.47475
age		1.043691	.0103505	4.31	0.000	1.0236 1.064176
tenure		1.024486	.0106404	2.33	0.020	1.003842 1.045554
Asian		5.70e-07	.0018811	-0.00	0.997	0 .
Black		7.95e-08	.0003862	-0.00	0.997	0 .
Latino		1.395964	1.01608	0.46	0.647	.335205 5.813507
Other		6.29e-07	.0009185	-0.01	0.992	0 .
AsianMale		.2015162	1671.221	-0.00	1.000	0 .
BlackMale		.3720007	2574.345	-0.00	1.000	0 .
LatinoMale		.2602679	.3263705	-1.07	0.283	.022286 3.039551
OtherMale		.1225657	452.6494	-0.00	1.000	0 .

-----> region = 17

```

Iteration 0: log likelihood = -1230.2952

```

```

Iteration 1: log likelihood = -1205.5449
Iteration 2: log likelihood = -1027.1536
Iteration 3: log likelihood = -1016.1236 (backed up)
Iteration 4: log likelihood = -1010.4169 (backed up)
Iteration 5: log likelihood = -1007.2539
Iteration 6: log likelihood = -1004.026
Iteration 7: log likelihood = -1003.9026
Iteration 8: log likelihood = -1003.8894
Iteration 9: log likelihood = -1003.8876
Iteration 10: log likelihood = -1003.8874
Iteration 11: log likelihood = -1003.8874
Iteration 12: log likelihood = -1003.8873

```

```

Multinomial logistic regression      Number of obs   =      6237
                                     LR chi2(22)        =      452.82
                                     Prob > chi2         =      0.0000
Log likelihood = -1003.8873          Pseudo R2       =      0.1840

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	10.35026	3.159003	7.66	0.000	5.690577	18.8255
	age	1.032203	.0168873	1.94	0.053	.9996293	1.065838
	tenure	1.0156	.0190453	0.83	0.409	.9789494	1.053622
	Asian	4.54e-06	.0154735	-0.00	0.997	0	.
	Black	3.45e-06	.003022	-0.01	0.989	0	.
	Latino	1.167374	1.209965	0.15	0.881	.1530921	8.901578
	Other	3.87e-06	.0048407	-0.01	0.992	0	.
	AsianMale	.1098801	603.2393	-0.00	1.000	0	.
	BlackMale	316448.1	2.78e+08	0.01	0.988	0	.
	LatinoMale	1.186898	1.384573	0.15	0.883	.1206276	11.67832
	OtherMale	712570.7	8.91e+08	0.01	0.991	0	.
3							
	male	11.83505	2.120013	13.79	0.000	8.330944	16.81302
	age	1.057864	.0105828	5.62	0.000	1.037324	1.07881
	tenure	1.043794	.010562	4.24	0.000	1.023297	1.064702
	Asian	7.05e-06	.0111392	-0.01	0.994	0	.
	Black	2.643768	1.443685	1.78	0.075	.9065813	7.709743
	Latino	1.903061	1.008453	1.21	0.225	.6735835	5.376678
	Other	2.206349	2.264695	0.77	0.441	.2950956	16.49626
	AsianMale	.1127803	315.7107	-0.00	0.999	0	.
	BlackMale	.1991167	.1870476	-1.72	0.086	.0315864	1.255206
	LatinoMale	.6918816	.435182	-0.59	0.558	.2016695	2.373687
	OtherMale	.4025248	.5985872	-0.61	0.541	.0218266	7.423329

-----> region = 18

```

Iteration 0: log likelihood = -1039.1286
Iteration 1: log likelihood = -924.54724
Iteration 2: log likelihood = -885.15912
Iteration 3: log likelihood = -883.98024
Iteration 4: log likelihood = -883.87484
Iteration 5: log likelihood = -883.85697
Iteration 6: log likelihood = -883.85281
Iteration 7: log likelihood = -883.85185
Iteration 8: log likelihood = -883.85165
Iteration 9: log likelihood = -883.85161

```

```

Multinomial logistic regression      Number of obs   =      5522
                                     LR chi2(22)        =      310.55
                                     Prob > chi2         =      0.0000

```

Log likelihood = -883.85161

Pseudo R2 = 0.1494

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
male	8.438477	3.16099	5.69	0.000	4.049584	17.584
age	1.014127	.0198997	0.71	0.475	.9758652	1.05389
tenure	1.053181	.0204034	2.67	0.007	1.01394	1.093939
Asian	5.62e-06	.014901	-0.00	0.996	0	.
Black	2.602379	2.756413	0.90	0.367	.3264275	20.74696
Latino	2.444679	1.618737	1.35	0.177	.6677305	8.950398
Other	6.33e-06	.0076111	-0.01	0.992	0	.
AsianMale	.2095386	1723.947	-0.00	1.000	0	.
BlackMale	1.115659	1.338914	0.09	0.927	.1061684	11.72378
LatinoMale	.8476822	.6537043	-0.21	0.830	.1869884	3.842832
OtherMale	.1448525	393.9153	-0.00	0.999	0	.
3						
male	7.515782	1.44658	10.48	0.000	5.153965	10.95991
age	1.050626	.0112245	4.62	0.000	1.028855	1.072858
tenure	1.040318	.0106359	3.87	0.000	1.01968	1.061375
Asian	4.61e-06	.0067679	-0.01	0.993	0	.
Black	1.270006	.9378571	0.32	0.746	.2986907	5.39995
Latino	2.159964	.7766987	2.14	0.032	1.067491	4.370479
Other	4.76e-06	.0031816	-0.02	0.985	0	.
AsianMale	.3144344	1462.708	-0.00	1.000	0	.
BlackMale	.3215007	.3372001	-1.08	0.279	.0411543	2.511591
LatinoMale	.4669414	.2167192	-1.64	0.101	.1880176	1.159648
OtherMale	.1726018	259.8152	-0.00	0.999	0	.

-----> region = 19

Iteration 0: log likelihood = -1848.9755  
 Iteration 1: log likelihood = -1791.6482  
 Iteration 2: log likelihood = -1613.9071  
 Iteration 3: log likelihood = -1609.4346  
 Iteration 4: log likelihood = -1609.1802  
 Iteration 5: log likelihood = -1609.1367  
 Iteration 6: log likelihood = -1609.1267  
 Iteration 7: log likelihood = -1609.1243  
 Iteration 8: log likelihood = -1609.1238  
 Iteration 9: log likelihood = -1609.1237  
 Iteration 10: log likelihood = -1609.1237

Multinomial logistic regression

Number of obs = 9322  
 LR chi2(22) = 479.70  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1297

Log likelihood = -1609.1237

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	2.212291	.4550925	3.86	0.000	1.478226	3.310882
	age	1.020349	.0098453	2.09	0.037	1.001234	1.039829
	tenure	1.080732	.0103329	8.12	0.000	1.060668	1.101175
	Asian	5.74e-07	.0011044	-0.01	0.994	0	.
	Black	1.339211	.6474864	0.60	0.546	.5191667	3.454545
	Latino	1.041333	.2167683	0.19	0.846	.6924683	1.565955
	Other	8.50e-07	.0009029	-0.01	0.989	0	.



AsianMale		3220847	6.20e+09	0.01	0.994	0	.
BlackMale		.5810349	.5122145	-0.62	0.538	.1032335	3.27027
LatinoMale		1.630581	.4728214	1.69	0.092	.9236731	2.8785
OtherMale		.5111999	1148.165	-0.00	1.000	0	.
-----							
3							
male		2.928333	.6609423	4.76	0.000	1.881478	4.557658
age		1.01982	.0121259	1.65	0.099	.9963288	1.043866
tenure		1.125828	.0126804	10.52	0.000	1.101247	1.150958
Asian		1.00e-06	.0015151	-0.01	0.993	0	.
Black		.6055099	.4516556	-0.67	0.501	.1403466	2.612406
Latino		1.209614	.2817853	0.82	0.414	.7662234	1.909582
Other		1.62e-06	.0014935	-0.01	0.988	0	.
AsianMale		.5552289	1289.883	-0.00	1.000	0	.
BlackMale		2.21668	2.155723	0.82	0.413	.3295406	14.91067
LatinoMale		.9092067	.298319	-0.29	0.772	.4779425	1.729616
OtherMale		1285673	1.18e+09	0.02	0.988	0	.
-----							

-----> region = 20

Iteration 0: log likelihood = -4001.2437  
Iteration 1: log likelihood = -3879.6749  
Iteration 2: log likelihood = -3578.3325  
Iteration 3: log likelihood = -3572.715  
Iteration 4: log likelihood = -3572.5901  
Iteration 5: log likelihood = -3572.5636  
Iteration 6: log likelihood = -3572.5572  
Iteration 7: log likelihood = -3572.5558  
Iteration 8: log likelihood = -3572.5555  
Iteration 9: log likelihood = -3572.5555  
Iteration 10: log likelihood = -3572.5554

Multinomial logistic regression                      Number of obs    =        19357  
   LR chi2(22)        =        857.38  
   Prob > chi2        =        0.0000  
Log likelihood = -3572.5554                              Pseudo R2        =        0.1071

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
male		2.756962	.3568461	7.84	0.000	2.139222 3.553085
age		1.0197	.0066259	3.00	0.003	1.006796 1.032769
tenure		1.056388	.007289	7.95	0.000	1.042198 1.070772
Asian		1.45e-06	.0011803	-0.02	0.987	0 .
Black		1.13379	.3242665	0.44	0.661	.6472721 1.985995
Latino		1.246553	.1979701	1.39	0.165	.9131207 1.701742
Other		.5910844	.5962264	-0.52	0.602	.0818546 4.268309
AsianMale		.2825129	661.4773	-0.00	1.000	0 .
BlackMale		1.828563	.7080376	1.56	0.119	.8560841 3.905742
LatinoMale		1.241145	.2774551	0.97	0.334	.800826 1.923564
OtherMale		3.110785	3.663182	0.96	0.335	.3093973 31.27688
-----						
3						
male		3.123239	.4100433	8.67	0.000	2.414642 4.039779
age		1.022137	.0073578	3.04	0.002	1.007818 1.036661
tenure		1.107128	.0077337	14.57	0.000	1.092074 1.12239
Asian		.9704595	.9954994	-0.03	0.977	.129961 7.246725
Black		1.053373	.2966076	0.18	0.853	.6065987 1.829206
Latino		1.370281	.2205057	1.96	0.050	.9996187 1.878387
Other		.7252442	.733399	-0.32	0.751	.0999336 5.263285
AsianMale		2.054696	3.071383	0.48	0.630	.1097391 38.47101
BlackMale		.9639395	.4188147	-0.08	0.933	.4113559 2.258821

LatinoMale		1.113475	.2494394	0.48	0.631	.7177866	1.727292
OtherMale		1.02e-06	.0009751	-0.01	0.988	0	.

### 1994-95

```
. mlogit admin male Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale
tenure age pyperstudecodis pyperstudlep pyperstud
> black pyperstudlatino if validcert1995a==1, rrr
```

```
Iteration 0: log likelihood = -20574.843
Iteration 1: log likelihood = -19679.973
Iteration 2: log likelihood = -19663.87
Iteration 3: log likelihood = -19663.83
Iteration 4: log likelihood = -19663.83
```

Multinomial logistic regression		Number of obs	=	19981
		LR chi2(30)	=	1822.03
		Prob > chi2	=	0.0000
Log likelihood = -19663.83		Pseudo R2	=	0.0443

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.400769	.0623329	7.57	0.000	1.283775	1.528425
	Asian	.49428	.2670504	-1.30	0.192	.1714291	1.425153
	Black	1.2838	.1014581	3.16	0.002	1.099581	1.498882
	Latino	1.312004	.099449	3.58	0.000	1.130875	1.522144
	Other	.8791098	.2697515	-0.42	0.675	.4817871	1.604098
	AsianMale	3.498102	2.417694	1.81	0.070	.9026701	13.55613
	BlackMale	1.707105	.1910345	4.78	0.000	1.370901	2.125761
	LatinoMale	1.414192	.1397044	3.51	0.000	1.165253	1.716312
	OtherMale	1.178251	.5175675	0.37	0.709	.4981182	2.787039
	tenure	1.001523	.0024504	0.62	0.534	.9967319	1.006337
	age	.9810666	.0025221	-7.44	0.000	.9761357	.9860224
pyperstude~s		.9916077	.0011507	-7.26	0.000	.989355	.9938656
pyperstudlep		1.011318	.0016971	6.71	0.000	1.007997	1.01465
pyperstudb~k		1.000294	.0012509	0.24	0.814	.9978455	1.002749
pyperstudl~o		1.000506	.0011665	0.43	0.664	.9982224	1.002795
3							
	male	1.925684	.079003	15.97	0.000	1.776903	2.086923
	Asian	.8552661	.3316812	-0.40	0.687	.3999413	1.828968
	Black	.9148681	.0777666	-1.05	0.295	.7744685	1.08072
	Latino	1.450261	.1080472	4.99	0.000	1.253227	1.678271
	Other	.3231089	.1414043	-2.58	0.010	.1370351	.7618436
	AsianMale	1.826568	.9989595	1.10	0.271	.6253304	5.335343
	BlackMale	1.185487	.1463033	1.38	0.168	.9307826	1.50989
	LatinoMale	.9083381	.089374	-0.98	0.329	.7490234	1.101538
	OtherMale	2.871593	1.521385	1.99	0.046	1.016604	8.111364
	tenure	1.025563	.0022949	11.28	0.000	1.021075	1.030071
	age	1.013527	.0025061	5.43	0.000	1.008627	1.018451
pyperstude~s		1.025502	.0011922	21.66	0.000	1.023168	1.027841
pyperstudlep		1.008257	.0016241	5.10	0.000	1.005078	1.011445
pyperstudb~k		.9765091	.0012698	-18.28	0.000	.9740236	.979001
pyperstudl~o		.9743351	.0011435	-22.15	0.000	.9720965	.9765788

```
. mlogtest, all
```

```
**** Likelihood-ratio tests for independent variables (N=19981)
```

```
Ho: All coefficients associated with given variable(s) are 0.
```

	chi2	df	P>chi2
male	266.012	2	0.000
Asian	2.012	2	0.366
Black	13.428	2	0.001
Latino	29.618	2	0.000
Other	8.775	2	0.012
AsianMale	3.878	2	0.144
BlackMale	22.981	2	0.000
LatinoMale	18.189	2	0.000
OtherMale	4.371	2	0.112
tenure	138.688	2	0.000
age	121.359	2	0.000
pyperstude~s	708.092	2	0.000
pyperstudlep	53.833	2	0.000
pyperstudb~k	395.800	2	0.000
pyperstudl~o	584.558	2	0.000

\*\*\*\* Wald tests for independent variables (N=19981)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
male	263.172	2	0.000
Asian	1.733	2	0.420
Black	13.654	2	0.001
Latino	29.755	2	0.000
Other	6.665	2	0.036
AsianMale	3.656	2	0.161
BlackMale	22.897	2	0.000
LatinoMale	18.153	2	0.000
OtherMale	3.968	2	0.138
tenure	136.746	2	0.000
age	119.774	2	0.000
pyperstude~s	662.423	2	0.000
pyperstudlep	53.404	2	0.000
pyperstudb~k	364.186	2	0.000
pyperstudl~o	541.853	2	0.000

\*\*\*\* Hausman tests of IIA assumption (N=19981)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-75.643	16	---	---
3	-12.338	16	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=19981)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-4704.631	-4693.975	21.311	16	0.167	for Ho
3	-4417.380	-4411.799	11.161	16	0.799	for Ho

\*\*\*\* Wald tests for combining alternatives (N=19981)

Ho: All coefficients except intercepts associated with a given pair

of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1021.129	15	0.000
2-	1	443.617	15	0.000
3-	1	1103.866	15	0.000

\*\*\*\* LR tests for combining alternatives (N=19981)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1121.995	15	0.000
2-	1	452.839	15	0.000
3-	1	1222.626	15	0.000

by region

```
. bysort region: mlogit admin male Asian Black Latino Other AsianMale BlackMale
LatinoMale OtherMale tenure age pyperstudecodis pyperst
> udlep pyperstudblack pyperstudlatino if validcert1995a==1, rrr
```

-----> region = 1

```
note: AsianMale omitted because of collinearity
note: BlackMale omitted because of collinearity
Iteration 0: log likelihood = -1331.6885
Iteration 1: log likelihood = -1261.042
Iteration 2: log likelihood = -1259.0501
Iteration 3: log likelihood = -1258.7574
Iteration 4: log likelihood = -1258.7093
Iteration 5: log likelihood = -1258.7064
Iteration 6: log likelihood = -1258.7057
Iteration 7: log likelihood = -1258.7056
Iteration 8: log likelihood = -1258.7055
Iteration 9: log likelihood = -1258.7055
```

```
Multinomial logistic regression      Number of obs   =      1252
LR chi2(26)                        =      145.97
Prob > chi2                         =      0.0000
Pseudo R2                          =      0.0548
Log likelihood = -1258.7055
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	male	1.674293	.5062673	1.70	0.088	.9256526 3.028413
	Asian	9.44e-07	.0014899	-0.01	0.993	0 .
	Black	1.787728	2.620982	0.40	0.692	.1010086 31.64061
	Latino	1.317062	.3314241	1.09	0.274	.8042861 2.156759
	Other	5.94e-07	.0013192	-0.01	0.995	0 .
	AsianMale	(omitted)				
	BlackMale	(omitted)				
	LatinoMale	1.458029	.4972195	1.11	0.269	.7472878 2.844751
	OtherMale	4159603	9.24e+09	0.01	0.995	0 .
	tenure	1.007452	.0095258	0.79	0.432	.9889535 1.026296
	age	.9769374	.0096684	-2.36	0.018	.9581704 .996072
	pyperstude~s	1.004324	.0062618	0.69	0.489	.9921262 1.016672
	pyperstudlep	1.003494	.0037636	0.93	0.352	.9961444 1.010898
	pyperstudb~k	.5641328	.2144281	-1.51	0.132	.2678168 1.188297

```

pyperstudl~o | .9704249 .0156147 -1.87 0.062 .9402982 1.001517
-----+-----
3
  male | .6047344 .2060258 -1.48 0.140 .310149 1.179123
  Asian | 4.40e-07 .000679 -0.01 0.992 0 .
  Black | 1.08e-06 .0014474 -0.01 0.992 0 .
  Latino | 1.127988 .2622756 0.52 0.604 .7151321 1.779192
  Other | 3.89e-07 .0008825 -0.01 0.995 0 .
  AsianMale | (omitted)
  BlackMale | (omitted)
  LatinoMale | 2.733261 1.032876 2.66 0.008 1.303225 5.732484
  OtherMale | 3.09e+07 7.02e+10 0.01 0.994 0 .
  tenure | 1.040771 .0097052 4.29 0.000 1.021922 1.059967
  age | 1.00698 .0104201 0.67 0.501 .9867624 1.027611
pyperstude~s | 1.02827 .0071705 4.00 0.000 1.014312 1.042421
pyperstudlep | 1.009701 .0038804 2.51 0.012 1.002124 1.017335
pyperstudb~k | .9210823 .112458 -0.67 0.501 .7250574 1.170104
pyperstudl~o | .9423763 .0149467 -3.74 0.000 .9135319 .9721314
-----+-----

```

```

-----> region = 2

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -805.14718
Iteration 1: log likelihood = -763.28873
Iteration 2: log likelihood = -762.31216
Iteration 3: log likelihood = -761.69432
Iteration 4: log likelihood = -761.68684
Iteration 5: log likelihood = -761.68508
Iteration 6: log likelihood = -761.68466
Iteration 7: log likelihood = -761.68457
Iteration 8: log likelihood = -761.68456

```

```

Multinomial logistic regression
Number of obs   =      855
LR chi2(26)     =      86.93
Prob > chi2     =      0.0000
Log likelihood = -761.68456
Pseudo R2      =      0.0540

```

```

-----+-----
      admin |      RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1
      |      (base outcome)
-----+-----
2
  male | 1.564344   .4403002     1.59   0.112   .9010586   2.715885
  Asian | (omitted)
  Black | 2.161312   1.327636     1.25   0.210   .6484061   7.204231
  Latino | .974127   .2834182    -0.09   0.928   .5507596   1.722936
  Other | 7822513   6.43e+09     0.02   0.985         0 .
  AsianMale | (omitted)
  BlackMale | 2.207217   2.463536     0.71   0.478   .2476256  19.67409
  LatinoMale | 1.380052   .5620743     0.79   0.429   .621175   3.066034
  OtherMale | 7.50e-07   .000617    -0.02   0.986         0 .
  tenure | 1.009575   .0144643     0.67   0.506   .9816199   1.038326
  age | .9733095   .0148601    -1.77   0.076   .9446157   1.002875
pyperstude~s | 1.00084   .006733     0.12   0.901   .9877306   1.014124
pyperstudlep | .9815485   .0173244    -1.06   0.291   .9481739   1.016098
pyperstudb~k | 1.012241   .019558     0.63   0.529   .9746245   1.051309
pyperstudl~o | 1.004241   .0071746     0.59   0.554   .9902773   1.018402
-----+-----
3
  male | 1.680177   .4112051     2.12   0.034   1.039998   2.714422
  Asian | (omitted)
  Black | .3941208   .4252862    -0.86   0.388   .0475465   3.266934
  Latino | 1.390318   .3455576     1.33   0.185   .8541841   2.262958

```

Other		2.728315	4032.941	0.00	0.999	0	.
AsianMale		(omitted)					
BlackMale		14.00043	19.52314	1.89	0.058	.9102657	215.3349
LatinoMale		1.122647	.3921642	0.33	0.741	.5661096	2.226311
OtherMale		2.42e-06	.0040224	-0.01	0.994	0	.
tenure		1.017368	.011845	1.48	0.139	.9944155	1.040851
age		1.029295	.0132792	2.24	0.025	1.003595	1.055654
pyperstude~s		1.024536	.0060143	4.13	0.000	1.012815	1.036392
pyperstudlep		.9926805	.0134922	-0.54	0.589	.9665854	1.01948
pyperstudb~k		.9600183	.0203906	-1.92	0.055	.9208739	1.000827
pyperstudl~o		.9783946	.0057099	-3.74	0.000	.9672672	.9896501

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -335.23824  
Iteration 1: log likelihood = -303.42121  
Iteration 2: log likelihood = -300.77339  
Iteration 3: log likelihood = -300.08501  
Iteration 4: log likelihood = -300.02205  
Iteration 5: log likelihood = -300.00647  
Iteration 6: log likelihood = -300.00333  
Iteration 7: log likelihood = -300.00265  
Iteration 8: log likelihood = -300.0025  
Iteration 9: log likelihood = -300.00246  
Iteration 10: log likelihood = -300.00246

Multinomial logistic regression	Number of obs	=	331
	LR chi2(24)	=	70.47
	Prob > chi2	=	0.0000
Log likelihood = -300.00246	Pseudo R2	=	0.1051

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		1.222337	.4884721	0.50	0.615	.5585119 2.675158
Asian		(omitted)				
Black		5.7588	4.778304	2.11	0.035	1.132547 29.28246
Latino		4.725587	3.929969	1.87	0.062	.9259035 24.11825
Other		2.83e-07	.0009449	-0.00	0.996	0 .
AsianMale		(omitted)				
BlackMale		.2377482	.3651033	-0.94	0.350	.0117206 4.822631
LatinoMale		.7543511	.8229565	-0.26	0.796	.0889116 6.400127
OtherMale		(omitted)				
tenure		.983977	.0221787	-0.72	0.474	.9414538 1.028421
age		.9827125	.0257869	-0.66	0.506	.9334488 1.034576
pyperstude~s		.9773094	.0151929	-1.48	0.140	.9479808 1.007545
pyperstudlep		1.041466	.0420727	1.01	0.315	.9621845 1.127279
pyperstudb~k		1.048255	.0237311	2.08	0.037	1.002759 1.095814
pyperstudl~o		1.036424	.015717	2.36	0.018	1.006072 1.067691
3						
male		2.543921	.7579367	3.13	0.002	1.418719 4.561534
Asian		(omitted)				
Black		4.916215	4.017558	1.95	0.051	.9908876 24.39144
Latino		1.07e-07	.000258	-0.01	0.995	0 .
Other		1.666549	2.392124	0.36	0.722	.1000055 27.77232
AsianMale		(omitted)				
BlackMale		3.19e-08	.0000666	-0.01	0.993	0 .
LatinoMale		3.44e+07	8.29e+10	0.01	0.994	0 .

OtherMale		(omitted)					
tenure		.9826928	.0158994	-1.08	0.281	.9520194	1.014354
age		1.040296	.0218699	1.88	0.060	.9983026	1.084055
pyperstude~s		1.024181	.0114869	2.13	0.033	1.001913	1.046944
pyperstudlep		1.042315	.0338396	1.28	0.202	.9780567	1.110795
pyperstudb~k		.9909118	.0164018	-0.55	0.581	.9592808	1.023586
pyperstudl~o		.9770516	.0100536	-2.26	0.024	.9575442	.9969564

-----> region = 4

Iteration 0: log likelihood = -4157.1179  
Iteration 1: log likelihood = -3916.3632  
Iteration 2: log likelihood = -3906.4162  
Iteration 3: log likelihood = -3906.3864  
Iteration 4: log likelihood = -3906.3864

Multinomial logistic regression		Number of obs	=	4102
		LR chi2(30)	=	501.46
		Prob > chi2	=	0.0000
Log likelihood = -3906.3864		Pseudo R2	=	0.0603

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		1.334733	.1262215	3.05	0.002	1.108917 1.606534
Asian		.9096129	.7526133	-0.11	0.909	.1797103 4.604054
Black		.8800312	.1163731	-0.97	0.334	.6791057 1.140404
Latino		1.417511	.2806729	1.76	0.078	.9615788 2.089622
Other		1.154599	.7310548	0.23	0.820	.3337935 3.993783
AsianMale		.5296551	.7303383	-0.46	0.645	.0355039 7.901508
BlackMale		1.671591	.3153135	2.72	0.006	1.154961 2.419317
LatinoMale		1.423911	.4361469	1.15	0.249	.7811921 2.59542
OtherMale		1.965605	2.044601	0.65	0.516	.2559089 15.09756
tenure		1.013914	.0056328	2.49	0.013	1.002933 1.025014
age		.9758718	.0053278	-4.47	0.000	.9654852 .9863701
pyperstude~s		.9998778	.0027038	-0.05	0.964	.9945925 1.005191
pyperstudlep		1.00048	.004904	0.10	0.922	.9909143 1.010138
pyperstudb~k		.990546	.0023544	-4.00	0.000	.9859422 .9951713
pyperstudl~o		.9971906	.0031666	-0.89	0.376	.9910034 1.003416
3						
male		1.400864	.1479147	3.19	0.001	1.138988 1.722951
Asian		.9620026	.8062019	-0.05	0.963	.1861361 4.971893
Black		.3951961	.066519	-5.52	0.000	.2841445 .5496497
Latino		2.455493	.4853566	4.54	0.000	1.666821 3.617334
Other		.4778739	.5160472	-0.68	0.494	.0575594 3.967438
AsianMale		.8627542	1.059454	-0.12	0.904	.0777356 9.575342
BlackMale		2.277613	.5205799	3.60	0.000	1.455211 3.564788
LatinoMale		.8591865	.2884702	-0.45	0.651	.4449367 1.659116
OtherMale		2.731603	4.463246	0.62	0.539	.1110746 67.17699
tenure		1.077055	.0066694	11.99	0.000	1.064062 1.090206
age		1.001912	.0063967	0.30	0.765	.9894526 1.014528
pyperstude~s		1.028069	.0032295	8.81	0.000	1.021759 1.034418
pyperstudlep		1.009169	.005537	1.66	0.096	.9983744 1.020079
pyperstudb~k		.9788429	.0028871	-7.25	0.000	.9732006 .984518
pyperstudl~o		.969798	.00384	-7.75	0.000	.962301 .9773535

-----> region = 5

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity

```

Iteration 0: log likelihood = -555.71035
Iteration 1: log likelihood = -496.83324
Iteration 2: log likelihood = -494.82667
Iteration 3: log likelihood = -494.66441
Iteration 4: log likelihood = -494.63014
Iteration 5: log likelihood = -494.62239
Iteration 6: log likelihood = -494.6212
Iteration 7: log likelihood = -494.62106
Iteration 8: log likelihood = -494.62103

```

```

Multinomial logistic regression      Number of obs   =      581
                                     LR chi2(26)        =     122.18
                                     Prob > chi2         =     0.0000
Log likelihood = -494.62103          Pseudo R2        =     0.1099

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	3.148305	.8933809	4.04	0.000	1.80524	5.490584
	Asian	1.390668	6115.738	0.00	1.000	0	.
	Black	1.832109	.7503629	1.48	0.139	.8209771	4.08857
	Latino	8.163086	10.70062	1.60	0.109	.6252484	106.5752
	Other	9.47e-07	.001024	-0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	.5461652	.3047621	-1.08	0.278	.1829576	1.630413
	LatinoMale	.0800183	.1458979	-1.39	0.166	.0022449	2.85227
	OtherMale	(omitted)					
	tenure	1.062476	.0179149	3.59	0.000	1.027937	1.098175
	age	.9512901	.0180857	-2.63	0.009	.9164952	.987406
pyperstude~s		.995317	.0086856	-0.54	0.591	.9784383	1.012487
pyperstudlep		.9578221	.0346742	-1.19	0.234	.8922168	1.028251
pyperstudb~k		1.001523	.0066345	0.23	0.818	.9886037	1.014611
pyperstudl~o		1.025105	.0241352	1.05	0.292	.9788756	1.073517
3							
	male	5.987787	1.603274	6.68	0.000	3.542835	10.12003
	Asian	2.13e+07	4.49e+10	0.01	0.994	0	.
	Black	1.999173	.8686776	1.59	0.111	.853076	4.685038
	Latino	.000019	.0133682	-0.02	0.988	0	.
	Other	1.34e-06	.0012639	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	.4609216	.2618451	-1.36	0.173	.1513797	1.403416
	LatinoMale	.0625222	62.46257	-0.00	0.998	0	.
	OtherMale	(omitted)					
	tenure	1.051123	.0149186	3.51	0.000	1.022286	1.080773
	age	1.00375	.0164345	0.23	0.819	.9720507	1.036484
pyperstude~s		1.038229	.0083221	4.68	0.000	1.022046	1.054669
pyperstudlep		.9679294	.03217	-0.98	0.327	.906887	1.03308
pyperstudb~k		.9721494	.0062754	-4.38	0.000	.9599273	.9845271
pyperstudl~o		.98772	.0207748	-0.59	0.557	.9478301	1.029289

-----> region = 6

```

Iteration 0: log likelihood = -692.35303
Iteration 1: log likelihood = -650.24211
Iteration 2: log likelihood = -648.64834
Iteration 3: log likelihood = -648.38862
Iteration 4: log likelihood = -648.3368
Iteration 5: log likelihood = -648.32491
Iteration 6: log likelihood = -648.32299
Iteration 7: log likelihood = -648.32252
Iteration 8: log likelihood = -648.32242

```



Iteration 9: log likelihood = -648.3224

Multinomial logistic regression

Number of obs = 666  
 LR chi2(30) = 88.06  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0636

Log likelihood = -648.3224

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.704839	.378949	2.40	0.016	1.102757	2.635647
	Asian	9.11e-07	.001685	-0.01	0.994	0	.
	Black	.6278747	.4973041	-0.59	0.557	.1329488	2.965252
	Latino	.262987	.3266455	-1.08	0.282	.0230511	3.000387
	Other	8.24e-07	.0015245	-0.01	0.994	0	.
	AsianMale	1.73e+13	5.56e+16	0.01	0.992	0	.
	BlackMale	7.372538	6.97301	2.11	0.035	1.154897	47.06421
	LatinoMale	17.73496	30.13012	1.69	0.091	.6349126	495.3889
	OtherMale	2.2684	4819.451	0.00	1.000	0	.
	tenure	.9900395	.0132849	-0.75	0.456	.964341	1.016423
	age	1.005071	.0139576	0.36	0.716	.9780832	1.032803
pyperstude~s		.9936953	.0075766	-0.83	0.407	.9789558	1.008657
pyperstudlep		1.033434	.0265598	1.28	0.201	.9826669	1.086823
pyperstudb~k		.9883373	.0099872	-1.16	0.246	.9689553	1.008107
pyperstudl~o		1.006146	.0133664	0.46	0.645	.9802868	1.032688
3							
	male	2.864507	.58846	5.12	0.000	1.915077	4.284631
	Asian	1.73e-06	.0028609	-0.01	0.994	0	.
	Black	1.132195	.650622	0.22	0.829	.3670912	3.491956
	Latino	1.23e-06	.000737	-0.02	0.982	0	.
	Other	1.06e-06	.0017563	-0.01	0.993	0	.
	AsianMale	580620.1	2.71e+09	0.00	0.998	0	.
	BlackMale	.8062677	.667838	-0.26	0.795	.1590092	4.088239
	LatinoMale	657317.3	3.93e+08	0.02	0.982	0	.
	OtherMale	2303987	3.81e+09	0.01	0.993	0	.
	tenure	1.006735	.0124857	0.54	0.588	.9825588	1.031507
	age	1.02471	.013469	1.86	0.063	.9986483	1.051452
pyperstude~s		1.02387	.0070963	3.40	0.001	1.010056	1.037873
pyperstudlep		1.020126	.0250456	0.81	0.417	.9722002	1.070415
pyperstudb~k		.9870241	.0085618	-1.51	0.132	.9703851	1.003948
pyperstudl~o		.9863189	.0118562	-1.15	0.252	.9633529	1.009832

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -968.78354

Iteration 1: log likelihood = -895.24257

Iteration 2: log likelihood = -887.76622

Iteration 3: log likelihood = -886.89841

Iteration 4: log likelihood = -886.87021

Iteration 5: log likelihood = -886.86308

Iteration 6: log likelihood = -886.86172

Iteration 7: log likelihood = -886.8615

Iteration 8: log likelihood = -886.86147

Multinomial logistic regression

Number of obs = 1018  
 LR chi2(28) = 163.84  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0846

Log likelihood = -886.86147

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	2.64432	.6458051	3.98	0.000	1.63844	4.267735
	Asian	1.44e+08	6.97e+11	0.00	0.997	0	.
	Black	.9761438	.4341163	-0.05	0.957	.4082842	2.333807
	Latino	1.585541	1.816657	0.40	0.687	.1678421	14.97801
	Other	5.27e-06	.0047365	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	4.043376	2.231038	2.53	0.011	1.371107	11.92386
	LatinoMale	1.789068	3.300545	0.32	0.753	.0481167	66.52083
	OtherMale	1963962	1.77e+09	0.02	0.987	0	.
	tenure	1.023218	.0140237	1.67	0.094	.9960978	1.051076
	age	.9787678	.0140052	-1.50	0.134	.9516995	1.006606
pyperstude~s		.9885836	.0079711	-1.42	0.154	.9730833	1.004331
pyperstudlep		.9774657	.032685	-0.68	0.495	.9154585	1.043673
pyperstudb~k		1.00948	.0074003	1.29	0.198	.995079	1.024089
pyperstudl~o		1.030801	.0160344	1.95	0.051	.9998486	1.062712
3							
	male	2.292618	.3856958	4.93	0.000	1.648659	3.188104
	Asian	1.793201	14729.12	0.00	1.000	0	.
	Black	.7731724	.2637268	-0.75	0.451	.3962178	1.508755
	Latino	.2793413	.3310131	-1.08	0.282	.027383	2.849639
	Other	5.55e-06	.0031619	-0.02	0.983	0	.
	AsianMale	(omitted)					
	BlackMale	2.046438	1.002206	1.46	0.144	.7836786	5.343911
	LatinoMale	3.204676	6.019952	0.62	0.535	.0806883	127.2793
	OtherMale	306272.1	1.74e+08	0.02	0.982	0	.
	tenure	1.006192	.0097856	0.63	0.526	.9871945	1.025556
	age	1.032135	.0108629	3.01	0.003	1.011062	1.053647
pyperstude~s		1.037701	.0061357	6.26	0.000	1.025745	1.049797
pyperstudlep		1.008366	.021278	0.39	0.693	.9675122	1.050944
pyperstudb~k		.9726693	.0055242	-4.88	0.000	.9619021	.9835571
pyperstudl~o		.993197	.012941	-0.52	0.600	.9681542	1.018888

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -308.18739  
Iteration 1: log likelihood = -277.01893  
Iteration 2: log likelihood = -275.16189  
Iteration 3: log likelihood = -274.9162  
Iteration 4: log likelihood = -274.87064  
Iteration 5: log likelihood = -274.86063  
Iteration 6: log likelihood = -274.85836  
Iteration 7: log likelihood = -274.85781  
Iteration 8: log likelihood = -274.8577  
Iteration 9: log likelihood = -274.85768

Multinomial logistic regression	Number of obs	=	315
	LR chi2(28)	=	66.66
	Prob > chi2	=	0.0001
Log likelihood = -274.85768	Pseudo R2	=	0.1081

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	3.40294	1.458665	2.86	0.004	1.468886	7.883526
	Asian	4.44e-07	.0014448	-0.00	0.996	0	.
	Black	3.00867	2.435954	1.36	0.174	.6154653	14.70772

Latino		6.14e-07	.0014105	-0.01	0.995	0	.
Other		9.44e-07	.0017218	-0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		1.43722	1.564581	0.33	0.739	.1701733	12.13822
LatinoMale		756734	4.21e+09	0.00	0.998	0	.
OtherMale		563783.5	3.04e+09	0.00	0.998	0	.
tenure		1.003822	.024104	0.16	0.874	.9576739	1.052195
age		1.006746	.0253452	0.27	0.789	.9582756	1.057667
pyperstude~s		.9700994	.0153644	-1.92	0.055	.9404484	1.000685
pyperstudlep		1.012234	.0809086	0.15	0.879	.8654537	1.183908
pyperstudb~k		1.028821	.016117	1.81	0.070	.9977127	1.0609
pyperstudl~o		1.015324	.0525673	0.29	0.769	.9173495	1.123763
-----							
3							
male		3.980262	1.196662	4.59	0.000	2.207998	7.175045
Asian		1.57e-07	.0003396	-0.01	0.994	0	.
Black		1.542588	1.179359	0.57	0.571	.3447294	6.902736
Latino		7.45e-07	.0011336	-0.01	0.993	0	.
Other		8.22e-07	.0010167	-0.01	0.991	0	.
AsianMale		(omitted)					
BlackMale		1.114927	1.173442	0.10	0.918	.1417006	8.772463
LatinoMale		8.76e+12	2.44e+16	0.01	0.991	0	.
OtherMale		7.13e+12	1.88e+16	0.01	0.991	0	.
tenure		1.013982	.0186935	0.75	0.451	.977997	1.05129
age		.996399	.0190178	-0.19	0.850	.9598133	1.034379
pyperstude~s		1.03056	.01081	2.87	0.004	1.009589	1.051966
pyperstudlep		.9914027	.0608509	-0.14	0.888	.8790317	1.118139
pyperstudb~k		.9832484	.0103363	-1.61	0.108	.9631969	1.003717
pyperstudl~o		.993997	.0384597	-0.16	0.876	.9214046	1.072308
-----							

-----> region = 9

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -199.43542  
Iteration 1: log likelihood = -173.91278  
Iteration 2: log likelihood = -171.75635  
Iteration 3: log likelihood = -171.58353  
Iteration 4: log likelihood = -171.55018  
Iteration 5: log likelihood = -171.54256  
Iteration 6: log likelihood = -171.54074  
Iteration 7: log likelihood = -171.54037  
Iteration 8: log likelihood = -171.54031  
Iteration 9: log likelihood = -171.5403

Multinomial logistic regression	Number of obs	=	215
	LR chi2(22)	=	55.79
	Prob > chi2	=	0.0001
Log likelihood = -171.5403	Pseudo R2	=	0.1399

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		2.518121	1.578385	1.47	0.141	.7371136 8.602386
Asian		(omitted)				
Black		4.469011	7.044647	0.95	0.342	.2034392 98.1721
Latino		17.11846	99433.28	0.00	1.000	0
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		2.82e-08	.0000767	-0.01	0.995	0

LatinoMale		3.52e-09	.0000267	-0.00	0.998	0	.
OtherMale		(omitted)					
tenure		1.040925	.0382437	1.09	0.275	.9686035	1.118646
age		.973697	.0400273	-0.65	0.517	.8983222	1.055396
pyperstude~s		.9449385	.0296598	-1.80	0.071	.8885584	1.004896
pyperstudlep		.6892917	.2215037	-1.16	0.247	.3671721	1.294006
pyperstudb~k		1.085224	.0390635	2.27	0.023	1.011299	1.164553
pyperstudl~o		1.044865	.0412633	1.11	0.266	.9670413	1.128952

-----

3							
male		3.418804	1.183605	3.55	0.000	1.734541	6.738512
Asian		(omitted)					
Black		4.837024	6.085648	1.25	0.210	.4108169	56.95188
Latino		2840679	6.43e+09	0.01	0.995	0	.
Other		(omitted)					
AsianMale		(omitted)					
BlackMale		3.89e-08	.0000501	-0.01	0.989	0	.
LatinoMale		5.14e-14	1.64e-10	-0.01	0.992	0	.
OtherMale		(omitted)					
tenure		1.019273	.020602	0.94	0.345	.9796826	1.060462
age		1.021249	.0232441	0.92	0.356	.9766925	1.067838
pyperstude~s		1.046477	.0142664	3.33	0.001	1.018886	1.074815
pyperstudlep		.8942719	.0863142	-1.16	0.247	.7401377	1.080505
pyperstudb~k		.9745255	.0178892	-1.41	0.160	.9400865	1.010226
pyperstudl~o		.997916	.0214636	-0.10	0.923	.9567226	1.040883

-----

-----> region = 10

Iteration 0: log likelihood = -2716.6146  
Iteration 1: log likelihood = -2577.2069  
Iteration 2: log likelihood = -2573.6384  
Iteration 3: log likelihood = -2572.9997  
Iteration 4: log likelihood = -2572.906  
Iteration 5: log likelihood = -2572.8952  
Iteration 6: log likelihood = -2572.893  
Iteration 7: log likelihood = -2572.8925  
Iteration 8: log likelihood = -2572.8924  
Iteration 9: log likelihood = -2572.8924

Multinomial logistic regression	Number of obs	=	2636
	LR chi2(30)	=	287.44
	Prob > chi2	=	0.0000
Log likelihood = -2572.8924	Pseudo R2	=	0.0529

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		1.743783	.2106744	4.60	0.000	1.376116 2.209683
Asian		1.01e-06	.0006272	-0.02	0.982	0 .
Black		1.600139	.295477	2.55	0.011	1.114237 2.297937
Latino		2.601877	.7064009	3.52	0.000	1.528228 4.42981
Other		.6403349	.4975708	-0.57	0.566	.1396313 2.93651
AsianMale		1477433	9.19e+08	0.02	0.982	0 .
BlackMale		1.464625	.3912308	1.43	0.153	.8676679 2.472291
LatinoMale		.5750072	.2314298	-1.37	0.169	.2612644 1.265512
OtherMale		9.02e-07	.0004808	-0.03	0.979	0 .
tenure		1.026022	.0070018	3.76	0.000	1.01239 1.039838
age		.9752989	.0067474	-3.62	0.000	.9621635 .9886137
pyperstude~s		.9939783	.0036998	-1.62	0.105	.9867533 1.001256
pyperstudlep		.9983987	.0073108	-0.22	0.827	.9841721 1.012831
pyperstudb~k		.9995545	.0033123	-0.13	0.893	.9930835 1.006068
pyperstudl~o		1.005484	.0048255	1.14	0.254	.996071 1.014987

-----+-----						
3						
	male	1.697925	.1964987	4.57	0.000	1.353349 2.130232
	Asian	.8341439	.5020002	-0.30	0.763	.2564338 2.713356
	Black	1.305232	.2553015	1.36	0.173	.8895979 1.915056
	Latino	2.96217	.8317976	3.87	0.000	1.708389 5.136097
	Other	.2139947	.2273243	-1.45	0.147	.0266797 1.716428
	AsianMale	3.884537	3.044439	1.73	0.083	.8360354 18.04903
	BlackMale	1.401534	.4083162	1.16	0.247	.7918061 2.480783
	LatinoMale	.9187694	.372358	-0.21	0.834	.4151761 2.033203
	OtherMale	4.125635	4.937474	1.18	0.236	.395176 43.07161
	tenure	1.052201	.0068929	7.77	0.000	1.038778 1.065798
	age	1.001648	.0068414	0.24	0.810	.9883283 1.015147
	pyperstude~s	1.024193	.0036127	6.78	0.000	1.017137 1.031299
	pyperstudlep	1.002546	.0073852	0.35	0.730	.9881752 1.017126
	pyperstudb~k	.9731917	.0032941	-8.03	0.000	.9667567 .9796694
	pyperstudl~o	.9674013	.0051035	-6.28	0.000	.9574501 .9774559
-----+-----						

-----> region = 11

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1920.6683  
Iteration 1: log likelihood = -1738.6853  
Iteration 2: log likelihood = -1735.2771  
Iteration 3: log likelihood = -1734.8986  
Iteration 4: log likelihood = -1734.8369  
Iteration 5: log likelihood = -1734.8232  
Iteration 6: log likelihood = -1734.8199  
Iteration 7: log likelihood = -1734.8192  
Iteration 8: log likelihood = -1734.819  
Iteration 9: log likelihood = -1734.819  
Iteration 10: log likelihood = -1734.819

Multinomial logistic regression	Number of obs	=	1796
	LR chi2(28)	=	371.70
	Prob > chi2	=	0.0000
Log likelihood = -1734.819	Pseudo R2	=	0.0968

-----+-----						
	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
1		(base outcome)				
-----+-----						
2						
	male	1.088828	.1473862	0.63	0.530	.8351001 1.419645
	Asian	3.45e-07	.0005286	-0.01	0.992	0 .
	Black	1.930503	.564212	2.25	0.024	1.088671 3.423293
	Latino	3.449232	1.398414	3.05	0.002	1.558195 7.635248
	Other	.1538101	.1644425	-1.75	0.080	.0189209 1.250338
	AsianMale	(omitted)				
	BlackMale	2.0647	.8591579	1.74	0.081	.913392 4.667205
	LatinoMale	.4826933	.271349	-1.30	0.195	.1603848 1.452712
	OtherMale	9.24e-07	.0014635	-0.01	0.993	0 .
	tenure	.9172551	.007891	-10.04	0.000	.9019186 .9328523
	age	.9912836	.0082412	-1.05	0.292	.975262 1.007568
	pyperstude~s	1.004126	.0045882	0.90	0.368	.9951733 1.013159
	pyperstudlep	1.005804	.0133154	0.44	0.662	.9800419 1.032243
	pyperstudb~k	.988443	.0040765	-2.82	0.005	.9804854 .9964651
	pyperstudl~o	.9835468	.0078036	-2.09	0.037	.9683704 .9989611
-----+-----						
3						
	male	1.746092	.2295036	4.24	0.000	1.349542 2.259164
	Asian	1.737177	2.128419	0.45	0.652	.1573755 19.17569
	Black	2.457771	.7669439	2.88	0.004	1.333292 4.53062
	Latino	3.195461	1.442743	2.57	0.010	1.318919 7.741927

```

      Other | 1.42e-07 .0001387 -0.02 0.987 0 .
AsianMale | (omitted)
BlackMale | .6916262 .3382604 -0.75 0.451 .2651954 1.803752
LatinoMale | .5902875 .3576365 -0.87 0.384 .1800311 1.93544
OtherMale | 1110059 1.09e+09 0.01 0.989 0 .
tenure | .944557 .0070937 -7.59 0.000 .9307553 .9585633
age | 1.056565 .0091447 6.36 0.000 1.038793 1.074641
pyperstude~s | 1.040386 .0046927 8.78 0.000 1.031229 1.049624
pyperstudlep | 1.007522 .0142154 0.53 0.595 .9800426 1.035773
pyperstudb~k | .9638588 .0042923 -8.27 0.000 .9554826 .9723084
pyperstudl~o | .9514987 .0084643 -5.59 0.000 .9350528 .9682339
-----

```

```

-----> region = 12

```

```

note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -722.65465
Iteration 1: log likelihood = -668.92378
Iteration 2: log likelihood = -666.51748
Iteration 3: log likelihood = -666.43058
Iteration 4: log likelihood = -666.40936
Iteration 5: log likelihood = -666.40538
Iteration 6: log likelihood = -666.40473
Iteration 7: log likelihood = -666.40457
Iteration 8: log likelihood = -666.40454
Iteration 9: log likelihood = -666.40453

```

```

Multinomial logistic regression          Number of obs   =      697
                                         LR chi2(26)      =     112.50
                                         Prob > chi2       =     0.0000
Log likelihood = -666.40453              Pseudo R2        =     0.0778

```

```

-----
      admin |          RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1          | (base outcome)
-----+-----
2          |
   male    | 1.301969   .313708     1.10  0.273     .811904   2.087838
   Asian    | 4.11e+07   1.45e+11     0.00  0.996           0           .
   Black    | 2.618056   1.171986     2.15  0.032     1.088766   6.295403
   Latino    | 2.377569   2.038918     1.01  0.313     .4427647   12.76713
   Other     | 1.87431    2.439178     0.48  0.629     .1462546   24.02
AsianMale   | (omitted)
BlackMale   | 1.025892   .6842154     0.04  0.969     .2775848   3.791472
LatinoMale   | 1.402801   1.799729     0.26  0.792     .1134863   17.33998
OtherMale    | (omitted)
   tenure   | .9962298   .0148243    -0.25  0.800     .9675944   1.025713
   age       | .9500587   .0147167    -3.31  0.001     .9216479   .9793453
pyperstude~s | .9806101   .0078835    -2.44  0.015     .9652799   .9961838
pyperstudlep | 1.047681   .0281447     1.73  0.083     .9939456   1.104322
pyperstudb~k | 1.031387   .0095188     3.35  0.001     1.012898   1.050213
pyperstudl~o | 1.008151   .0129327     0.63  0.527     .9831197   1.03382
-----
3          |
   male     | 2.313204   .4474283     4.34  0.000     1.583331   3.379531
   Asian     | 1.086522   5977.474     0.00  1.000           0           .
   Black     | 2.199592   1.005546     1.72  0.085     .8978721   5.388525
   Latino     | 1.88641    1.79592     0.67  0.505     .2919214   12.19007
   Other      | 3.62e-06   .0028456    -0.02  0.987           0           .
AsianMale    | (omitted)
BlackMale    | .4681612   .3194658    -1.11  0.266     .1228992   1.783372
LatinoMale    | .7669893   1.06764    -0.19  0.849     .0501103   11.73956
OtherMale     | (omitted)
   tenure     | 1.018576   .0118836     1.58  0.115     .9955484   1.042135

```

age		.9877139	.0122218	-1.00	0.318	.9640478	1.011961
pyperstude~s		1.029739	.0066736	4.52	0.000	1.016742	1.042903
pyperstudlep		.9791052	.0256671	-0.81	0.421	.930069	1.030727
pyperstudb~k		.9745072	.0069926	-3.60	0.000	.9608978	.9883093
pyperstudl~o		.9824638	.0104344	-1.67	0.096	.9622241	1.003129

-----> region = 13

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1133.7961  
Iteration 1: log likelihood = -1079.1122  
Iteration 2: log likelihood = -1077.6906  
Iteration 3: log likelihood = -1077.5216  
Iteration 4: log likelihood = -1077.493  
Iteration 5: log likelihood = -1077.4865  
Iteration 6: log likelihood = -1077.485  
Iteration 7: log likelihood = -1077.4847  
Iteration 8: log likelihood = -1077.4846  
Iteration 9: log likelihood = -1077.4846

Multinomial logistic regression	Number of obs	=	1049
	LR chi2(26)	=	112.62
	Prob > chi2	=	0.0000
Log likelihood = -1077.4846	Pseudo R2	=	0.0497

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		1.11973	.2054234	0.62	0.538	.7815442 1.604253
Asian		(omitted)				
Black		2.301357	.9058295	2.12	0.034	1.064009 4.97763
Latino		2.584381	.7860934	3.12	0.002	1.42379 4.69102
Other		1.27e-06	.0010424	-0.02	0.987	0 .
AsianMale		(omitted)				
BlackMale		3.650705	2.742357	1.72	0.085	.8374647 15.91428
LatinoMale		1.148663	.545045	0.29	0.770	.4532047 2.911327
OtherMale		521840.4	4.28e+08	0.02	0.987	0 .
tenure		1.026855	.0119514	2.28	0.023	1.003696 1.050549
age		.9771236	.0108343	-2.09	0.037	.9561179 .9985908
pyperstude~s		.9878177	.0065986	-1.83	0.067	.974969 1.000836
pyperstudlep		.9921557	.0155302	-0.50	0.615	.9621792 1.023066
pyperstudb~k		1.014445	.0079571	1.83	0.067	.9989685 1.030161
pyperstudl~o		1.016381	.0071547	2.31	0.021	1.002454 1.030501
3						
male		1.652626	.2780872	2.99	0.003	1.188346 2.298297
Asian		(omitted)				
Black		1.907042	.8108199	1.52	0.129	.8288073 4.388002
Latino		1.56753	.5348272	1.32	0.188	.8031439 3.059414
Other		.64941	.7602975	-0.37	0.712	.0654597 6.442644
AsianMale		(omitted)				
BlackMale		2.922469	2.255962	1.39	0.165	.6436888 13.26857
LatinoMale		1.048198	.5398125	0.09	0.927	.3820162 2.876103
OtherMale		1.83e-06	.0016448	-0.01	0.988	0 .
tenure		1.035503	.0116172	3.11	0.002	1.012982 1.058524
age		1.001129	.0104768	0.11	0.914	.9808037 1.021875
pyperstude~s		1.022029	.0063483	3.51	0.000	1.009662 1.034548
pyperstudlep		.9944049	.01537	-0.36	0.717	.964732 1.024991
pyperstudb~k		.980576	.0077406	-2.48	0.013	.9655215 .9958652
pyperstudl~o		.9798916	.0068501	-2.91	0.004	.9665573 .99341

-----> region = 14

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -271.11212  
 Iteration 1: log likelihood = -234.76426  
 Iteration 2: log likelihood = -230.23075  
 Iteration 3: log likelihood = -229.15619  
 Iteration 4: log likelihood = -229.05342  
 Iteration 5: log likelihood = -229.03066  
 Iteration 6: log likelihood = -229.02506  
 Iteration 7: log likelihood = -229.02392  
 Iteration 8: log likelihood = -229.02368  
 Iteration 9: log likelihood = -229.02362  
 Iteration 10: log likelihood = -229.02361

Multinomial logistic regression	Number of obs	=	283
	LR chi2 (26)	=	84.18
	Prob > chi2	=	0.0000
Log likelihood = -229.02361	Pseudo R2	=	0.1552

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.880217	.8619677	1.38	0.168	.765566	4.617781
	Asian	(omitted)					
	Black	5.03061	8.030908	1.01	0.312	.2201685	114.9439
	Latino	1.87e-06	.0040266	-0.01	0.995	0	.
	Other	5.30e-06	.0092003	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.1930017	382.5804	-0.00	0.999	0	.
	LatinoMale	2418432	5.21e+09	0.01	0.995	0	.
	OtherMale	1.717658	3998.731	0.00	1.000	0	.
	tenure	.9801344	.0302515	-0.65	0.516	.9226003	1.041256
	age	.9475717	.0332399	-1.54	0.125	.8846119	1.015013
pyperstude~s		.9304195	.0207998	-3.23	0.001	.8905328	.9720928
pyperstudlep		1.078895	.1578808	0.52	0.604	.8098746	1.437276
pyperstudb~k		1.158695	.0635896	2.68	0.007	1.040531	1.290278
pyperstudl~o		1.042166	.0249458	1.73	0.084	.9944022	1.092224
3							
	male	2.81966	.8434893	3.47	0.001	1.568785	5.067924
	Asian	(omitted)					
	Black	5.47e-07	.0019363	-0.00	0.997	0	.
	Latino	3.32e-07	.0003899	-0.01	0.990	0	.
	Other	1.324731	1.975423	0.19	0.850	.0712568	24.628
	AsianMale	(omitted)					
	BlackMale	8.02e+12	2.95e+16	0.01	0.994	0	.
	LatinoMale	2475118	2.90e+09	0.01	0.990	0	.
	OtherMale	.3372881	.704194	-0.52	0.603	.0056345	20.19061
	tenure	.9988189	.0178874	-0.07	0.947	.9643683	1.0345
	age	.994836	.0206411	-0.25	0.803	.9551918	1.036126
pyperstude~s		1.029552	.0108172	2.77	0.006	1.008567	1.050973
pyperstudlep		1.132349	.0710322	1.98	0.048	1.001347	1.28049
pyperstudb~k		.9075944	.0292038	-3.01	0.003	.8521235	.9666762
pyperstudl~o		.980636	.0141438	-1.36	0.175	.9533029	1.008753

-----> region = 15

note: Black omitted because of collinearity



note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -293.75171  
Iteration 1: log likelihood = -260.79714  
Iteration 2: log likelihood = -259.06898  
Iteration 3: log likelihood = -259.01509  
Iteration 4: log likelihood = -259.00317  
Iteration 5: log likelihood = -259.00025  
Iteration 6: log likelihood = -258.99968  
Iteration 7: log likelihood = -258.99959  
Iteration 8: log likelihood = -258.99957

Multinomial logistic regression                      Number of obs       =       298  
   LR chi2(22)       =       69.50  
   Prob > chi2       =       0.0000  
Log likelihood = -258.99957                      Pseudo R2       =       0.1183

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	.8522178	.3948282	-0.35	0.730	.3437108	2.113041
	Asian	1.23e-06	.0016841	-0.01	0.992	0	.
	Black	(omitted)					
	Latino	.8059995	.6396899	-0.27	0.786	.1701256	3.818561
	Other	7.245648	13055.5	0.00	0.999	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.445147	1.287348	0.41	0.679	.2521477	8.282646
	OtherMale	(omitted)					
	tenure	1.027787	.0269847	1.04	0.297	.9762352	1.08206
	age	.9615412	.0291136	-1.30	0.195	.9061397	1.02033
pyperstude~s		.9725992	.0112045	-2.41	0.016	.9508849	.9948094
pyperstudlep		1.013109	.0244841	0.54	0.590	.96624	1.062252
pyperstudb~k		1.09138	.0588592	1.62	0.105	.9819062	1.21306
pyperstudl~o		1.035192	.0134591	2.66	0.008	1.009146	1.06191
3							
	male	3.843046	1.223913	4.23	0.000	2.058682	7.174006
	Asian	.0000121	.0095996	-0.01	0.989	0	.
	Black	(omitted)					
	Latino	.6692581	.4902751	-0.55	0.584	.159233	2.812899
	Other	326525.1	2.72e+08	0.02	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.095289	.8674069	0.11	0.909	.2319663	5.171688
	OtherMale	(omitted)					
	tenure	.9993253	.0176135	-0.04	0.969	.965393	1.03445
	age	1.028975	.0195843	1.50	0.133	.9912971	1.068084
pyperstude~s		1.024872	.0082225	3.06	0.002	1.008882	1.041115
pyperstudlep		1.034279	.0189553	1.84	0.066	.9977861	1.072106
pyperstudb~k		.9938565	.0378823	-0.16	0.872	.9223142	1.070948
pyperstudl~o		.974905	.0086693	-2.86	0.004	.9580606	.9920456

-----> region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -427.53774  
Iteration 1: log likelihood = -396.30457  
Iteration 2: log likelihood = -393.36704  
Iteration 3: log likelihood = -392.89009

```

Iteration 4: log likelihood = -392.78738
Iteration 5: log likelihood = -392.76572
Iteration 6: log likelihood = -392.76128
Iteration 7: log likelihood = -392.76027
Iteration 8: log likelihood = -392.76002
Iteration 9: log likelihood = -392.75997
Iteration 10: log likelihood = -392.75996

```

Multinomial logistic regression

```

Number of obs   =      418
LR chi2(26)     =      69.56
Prob > chi2     =      0.0000
Pseudo R2      =      0.0813

```

Log likelihood = -392.75996

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.052016	.3375776	0.16	0.874	.5608976	1.973156
	Asian	(omitted)					
	Black	1.065906	1.057463	0.06	0.949	.1524966	7.450368
	Latino	.8469466	1.00675	-0.14	0.889	.082423	8.702893
	Other	3.70e-07	.0007802	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.7105377	1.143733	-0.21	0.832	.0302992	16.66261
	LatinoMale	5.228983	7.934632	1.09	0.276	.2671626	102.3432
	OtherMale	1.47e+07	3.11e+10	0.01	0.994	0	.
	tenure	1.046686	.0222219	2.15	0.032	1.004026	1.091159
	age	.9617933	.0218098	-1.72	0.086	.919983	1.005504
pyperstude~s		.9570438	.0123925	-3.39	0.001	.9330606	.9816434
pyperstudlep		.9865412	.0326579	-0.41	0.682	.9245651	1.052672
pyperstudb~k		1.045493	.022916	2.03	0.042	1.00153	1.091387
pyperstudl~o		1.033032	.014281	2.35	0.019	1.005418	1.061405
3							
	male	1.745936	.4174884	2.33	0.020	1.09267	2.789766
	Asian	(omitted)					
	Black	3.12e-07	.0002679	-0.02	0.986	0	.
	Latino	.2831229	.2606701	-1.37	0.171	.0465883	1.720572
	Other	2.78e-07	.0003886	-0.01	0.991	0	.
	AsianMale	(omitted)					
	BlackMale	.9031431	1316.716	-0.00	1.000	0	.
	LatinoMale	1.535038	2.358926	0.28	0.780	.0755196	31.20172
	OtherMale	2569312	3.60e+09	0.01	0.992	0	.
	tenure	1.037376	.0152782	2.49	0.013	1.007859	1.067757
	age	.9927454	.0155155	-0.47	0.641	.9627966	1.023626
pyperstude~s		1.012729	.0079356	1.61	0.106	.997294	1.028402
pyperstudlep		1.044093	.0226883	1.99	0.047	1.000558	1.089521
pyperstudb~k		.9936574	.0168056	-0.38	0.707	.961259	1.027148
pyperstudl~o		.9775906	.0097354	-2.28	0.023	.9586946	.996859

-----> region = 17

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -414.28559
Iteration 1: log likelihood = -377.1992
Iteration 2: log likelihood = -374.51524
Iteration 3: log likelihood = -374.37131
Iteration 4: log likelihood = -374.35039
Iteration 5: log likelihood = -374.3464
Iteration 6: log likelihood = -374.34554
Iteration 7: log likelihood = -374.34533
Iteration 8: log likelihood = -374.34529

```

Iteration 9: log likelihood = -374.34528

Multinomial logistic regression	Number of obs	=	414
	LR chi2(26)	=	79.88
	Prob > chi2	=	0.0000
Log likelihood = -374.34528	Pseudo R2	=	0.0964

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	male	.519969	.1281871	-2.65	0.008	.3207245	.8429906
	Asian	(omitted)					
	Black	.2081968	.2498667	-1.31	0.191	.019811	2.187967
	Latino	1.340269	.9375854	0.42	0.675	.3401994	5.280198
	Other	2.03e-07	.0003954	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	3.34944	6.204051	0.65	0.514	.0887812	126.364
	LatinoMale	.4255603	.3819765	-0.95	0.341	.0732716	2.471649
	OtherMale	2.02e+13	5.39e+16	0.01	0.991	0	.
	tenure	.9997406	.0154838	-0.02	0.987	.9698489	1.030554
	age	.9708324	.0170184	-1.69	0.091	.9380433	1.004768
	pypersude~s	.9642311	.0108533	-3.24	0.001	.9431919	.9857396
	pypersudlep	.9759938	.0221456	-1.07	0.284	.9335403	1.020378
	pypersudb~k	1.033218	.0153677	2.20	0.028	1.003532	1.063781
	pypersudl~o	1.035968	.011929	3.07	0.002	1.012849	1.059614
2							
	male	1.511556	.6181384	1.01	0.312	.6781605	3.369117
	Asian	(omitted)					
	Black	.5757824	.7934339	-0.40	0.689	.0386629	8.574773
	Latino	1.728278	2.134582	0.44	0.658	.1535683	19.45027
	Other	8.91e-07	.0029913	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	1.305227	2.538131	0.14	0.891	.0288701	59.00973
	LatinoMale	.6571583	.9242418	-0.30	0.765	.041737	10.3471
	OtherMale	8.70e+12	3.32e+16	0.01	0.994	0	.
	tenure	.9624434	.0227226	-1.62	0.105	.9189227	1.008025
	age	.9804633	.0242172	-0.80	0.424	.934129	1.029096
	pypersude~s	.9119574	.0164958	-5.10	0.000	.8801927	.9448684
	pypersudlep	1.01535	.0347904	0.44	0.657	.9494012	1.08588
	pypersudb~k	1.114373	.0232814	5.18	0.000	1.069664	1.160951
	pypersudl~o	1.066385	.0192218	3.57	0.000	1.029369	1.104733
3		(base outcome)					

-----> region = 18

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -391.93471  
Iteration 1: log likelihood = -361.26034  
Iteration 2: log likelihood = -358.78667  
Iteration 3: log likelihood = -358.62787  
Iteration 4: log likelihood = -358.61638  
Iteration 5: log likelihood = -358.61434  
Iteration 6: log likelihood = -358.61389  
Iteration 7: log likelihood = -358.61378  
Iteration 8: log likelihood = -358.61376

Multinomial logistic regression	Number of obs	=	395
	LR chi2(24)	=	66.64
	Prob > chi2	=	0.0000
Log likelihood = -358.61376	Pseudo R2	=	0.0850

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.307543	.5226908	0.67	0.502	.5972932	2.862359
	Asian	(omitted)					
	Black	7.926964	11.70445	1.40	0.161	.4388104	143.198
	Latino	3.944258	2.664692	2.03	0.042	1.049311	14.82609
	Other	6.29e-06	.0055252	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	1.526066	2.900441	0.22	0.824	.0367943	63.29462
	LatinoMale	1.20751	1.004994	0.23	0.821	.2362912	6.170697
	OtherMale	(omitted)					
	tenure	.9867814	.0217636	-0.60	0.546	.9450344	1.030373
	age	.9881478	.0272404	-0.43	0.665	.9361743	1.043007
pyperstude~s		.9836727	.014667	-1.10	0.270	.9553419	1.012844
pyperstudlep		.9801962	.024217	-0.81	0.418	.9338626	1.028829
pyperstudb~k		1.030872	.0265023	1.18	0.237	.9802152	1.084146
pyperstudl~o		.9960438	.0135979	-0.29	0.772	.9697457	1.023055
3							
	male	1.781155	.4729232	2.17	0.030	1.05851	2.997149
	Asian	(omitted)					
	Black	2.377436	3.018477	0.68	0.495	.1974211	28.63019
	Latino	1.660555	.8862032	0.95	0.342	.5834249	4.726303
	Other	7.43e-06	.0038882	-0.02	0.982	0	.
	AsianMale	(omitted)					
	BlackMale	1.084562	1.931832	0.05	0.964	.0330435	35.59774
	LatinoMale	.6678302	.4550477	-0.59	0.554	.1756598	2.538983
	OtherMale	(omitted)					
	tenure	1.005655	.0145857	0.39	0.697	.97747	1.034653
	age	1.021082	.0186457	1.14	0.253	.985183	1.058288
pyperstude~s		1.050266	.0112238	4.59	0.000	1.028496	1.072496
pyperstudlep		.9764444	.0126091	-1.85	0.065	.9520412	1.001473
pyperstudb~k		.9630472	.0217413	-1.67	0.095	.921364	1.006616
pyperstudl~o		.975595	.009715	-2.48	0.013	.9567386	.9948231

-----> region = 19

Iteration 0: log likelihood = -655.29046  
Iteration 1: log likelihood = -621.65344  
Iteration 2: log likelihood = -620.90557  
Iteration 3: log likelihood = -620.80125  
Iteration 4: log likelihood = -620.7848  
Iteration 5: log likelihood = -620.78086  
Iteration 6: log likelihood = -620.78002  
Iteration 7: log likelihood = -620.77984  
Iteration 8: log likelihood = -620.7798  
Iteration 9: log likelihood = -620.77979

Multinomial logistic regression	Number of obs	=	612
	LR chi2(30)	=	69.02
	Prob > chi2	=	0.0001
Log likelihood = -620.77979	Pseudo R2	=	0.0527

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	male	1.323035	.3666509	1.01	0.312	.7685623	2.277528

Asian		5419644	1.06e+10	0.01	0.994	0	.
Black		1.529801	1.022388	0.64	0.525	.4128238	5.668982
Latino		1.274755	.3624052	0.85	0.393	.7301885	2.225452
Other		1.913545	2.782209	0.45	0.655	.110721	33.07103
AsianMale		.7125773	1979.229	-0.00	1.000	0	.
BlackMale		.4472491	.4493776	-0.80	0.423	.0624161	3.204808
LatinoMale		1.149895	.452543	0.35	0.723	.5317001	2.486852
OtherMale		.3039449	1342.968	-0.00	1.000	0	.
tenure		1.035481	.0124951	2.89	0.004	1.011278	1.060263
age		.9751151	.0132898	-1.85	0.064	.9494124	1.001514
pyperstude~s		.994867	.0053837	-0.95	0.342	.9843709	1.005475
pyperstudlep		1.012113	.0067142	1.81	0.070	.9990383	1.025358
pyperstudb~k		1.013281	.0308181	0.43	0.664	.9546438	1.07552
pyperstudl~o		1.002337	.0097108	0.24	0.810	.9834838	1.021552
-----							
3							
male		1.277871	.3996907	0.78	0.433	.6922281	2.358982
Asian		1.142375	3758.99	0.00	1.000	0	.
Black		.2303713	.2617017	-1.29	0.196	.0248579	2.134976
Latino		1.272488	.3973819	0.77	0.440	.6899762	2.346784
Other		2.05e-06	.0024626	-0.01	0.991	0	.
AsianMale		.5328001	2479.371	-0.00	1.000	0	.
BlackMale		2.534456	3.654823	0.64	0.519	.1501088	42.79209
LatinoMale		.935281	.4160509	-0.15	0.880	.3911048	2.236614
OtherMale		2.10e+12	6.63e+15	0.01	0.993	0	.
tenure		1.075621	.0151121	5.19	0.000	1.046406	1.105652
age		.9879839	.0157265	-0.76	0.448	.9576364	1.019293
pyperstude~s		.9976435	.0064539	-0.36	0.715	.9850739	1.010374
pyperstudlep		1.01947	.0076303	2.58	0.010	1.004624	1.034536
pyperstudb~k		1.023406	.0345946	0.68	0.494	.9577988	1.093507
pyperstudl~o		1.000193	.0113536	0.02	0.986	.9781862	1.022695
-----							

-----> region = 20

```
Iteration 0: log likelihood = -1878.6351
Iteration 1: log likelihood = -1778.2369
Iteration 2: log likelihood = -1774.4591
Iteration 3: log likelihood = -1774.2731
Iteration 4: log likelihood = -1774.2281
Iteration 5: log likelihood = -1774.2195
Iteration 6: log likelihood = -1774.218
Iteration 7: log likelihood = -1774.2177
Iteration 8: log likelihood = -1774.2176
Iteration 9: log likelihood = -1774.2176
```

Multinomial logistic regression	Number of obs	=	1983
	LR chi2(30)	=	208.84
	Prob > chi2	=	0.0000
Log likelihood = -1774.2176	Pseudo R2	=	0.0556

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
male		1.642745	.2503813	3.26	0.001	1.218521 2.21466
Asian		1.92e-06	.0013713	-0.02	0.985	0 .
Black		1.293885	.4314529	0.77	0.440	.6730662 2.48733
Latino		1.01614	.1953477	0.08	0.934	.6971307 1.481128
Other		2.981485	2.748765	1.18	0.236	.4894085 18.16326
AsianMale		400644.3	2.55e+09	0.00	0.998	0 .
BlackMale		1.715954	.7913796	1.17	0.242	.6949294 4.237119
LatinoMale		1.58218	.4236728	1.71	0.087	.9361016 2.674169
OtherMale		1.37068	1.789683	0.24	0.809	.1060549 17.71501

tenure		1.022034	.0086683	2.57	0.010	1.005185	1.039166
age		.9888183	.0086262	-1.29	0.197	.972055	1.005871
pyperstude~s		1.001369	.0041782	0.33	0.743	.9932129	1.009592
pyperstudlep		1.000348	.0061934	0.06	0.955	.9882823	1.012561
pyperstudb~k		.9911594	.007379	-1.19	0.233	.9768018	1.005728
pyperstudl~o		.9917915	.0047928	-1.71	0.088	.9824422	1.00123
-----							
3							
male		1.71901	.2778938	3.35	0.001	1.252201	2.359843
Asian		.7671174	.8780918	-0.23	0.817	.0813808	7.231057
Black		1.067924	.3464718	0.20	0.839	.5654326	2.016972
Latino		1.046214	.2002909	0.24	0.813	.7188915	1.522571
Other		2.076	2.430575	0.62	0.533	.2092392	20.59737
AsianMale		2.93e+07	9.26e+10	0.01	0.996	0	.
BlackMale		.877207	.4406021	-0.26	0.794	.3277645	2.347698
LatinoMale		1.386813	.3760864	1.21	0.228	.8150472	2.359681
OtherMale		9.88e-07	.001066	-0.01	0.990	0	.
tenure		1.069834	.0092706	7.79	0.000	1.051818	1.088159
age		.9958113	.0092751	-0.45	0.652	.9777973	1.014157
pyperstude~s		1.027976	.0049625	5.72	0.000	1.018295	1.037748
pyperstudlep		1.019673	.0054081	3.67	0.000	1.009129	1.030328
pyperstudb~k		.9735302	.0075844	-3.44	0.001	.9587779	.9885095
pyperstudl~o		.9669541	.0053322	-6.09	0.000	.9565595	.9774617
-----							

-----> region = .

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -56.484205  
Iteration 1: log likelihood = -41.603153  
Iteration 2: log likelihood = -38.687575  
Iteration 3: log likelihood = -38.313203  
Iteration 4: log likelihood = -38.244554  
Iteration 5: log likelihood = -38.228678  
Iteration 6: log likelihood = -38.22491  
Iteration 7: log likelihood = -38.224148  
Iteration 8: log likelihood = -38.224025  
Iteration 9: log likelihood = -38.223995  
Iteration 10: log likelihood = -38.223988  
Iteration 11: log likelihood = -38.223987

Multinomial logistic regression	Number of obs	=	65
	LR chi2(24)	=	36.52
	Prob > chi2	=	0.0488
Log likelihood = -38.223987	Pseudo R2	=	0.3233

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
male		3.78215	4.722234	1.07	0.287	.3273137 43.7032
Asian		1.32e-07	.0012279	-0.00	0.999	0 .
Black		1.80e-08	.0000463	-0.01	0.994	0 .
Latino		4.87e-08	.000238	-0.00	0.997	0 .
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		2.85e+08	7.35e+11	0.01	0.994	0 .
LatinoMale		8.21e+07	4.02e+11	0.00	0.997	0 .
OtherMale		(omitted)				
tenure		.9771957	.0744714	-0.30	0.762	.8416123 1.134622
age		.8848976	.0740737	-1.46	0.144	.7510001 1.042668
pyperstude~s		.9816133	.0291284	-0.63	0.532	.9261512 1.040397

```

pyperstudlep | .9491633 .0579771 -0.85 0.393 .8420688 1.069878
pyperstudb~k | 1.07128 .0403899 1.83 0.068 .9949714 1.153441
pyperstudl~o | 1.044087 .027479 1.64 0.101 .9915947 1.099358
-----
3
    male | .7235273 .7910521 -0.30 0.767 .0848815 6.167323
    Asian | 2.71e-07 .002095 -0.00 0.998 0 .
    Black | .3802953 .555984 -0.66 0.508 .0216612 6.676675
    Latino | 1.33e-07 .0005226 -0.00 0.997 0 .
    Other | (omitted)
    AsianMale | (omitted)
    BlackMale | 2.94e-07 .0012694 -0.00 0.997 0 .
    LatinoMale | 3.43e+07 1.35e+11 0.00 0.996 0 .
    OtherMale | (omitted)
    tenure | 1.136419 .0615564 2.36 0.018 1.021954 1.263705
    age | 1.02726 .0562382 0.49 0.623 .9227424 1.143615
pyperstude~s | 1.009108 .021769 0.42 0.674 .9673314 1.05269
pyperstudlep | .9348442 .0593974 -1.06 0.289 .8253845 1.05882
pyperstudb~k | .9965949 .0263606 -0.13 0.897 .9462455 1.049623
pyperstudl~o | .9997663 .0230172 -0.01 0.992 .9556562 1.045912
-----

```

### 1995-96

```

. mlogit admin tenure exper male age pyperstudwhite pyperstudblack pyperstudlatino
pyperstudecodis pyperstudlep pyperstudmobile Asian Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert1996==1, rrr

```

```

Iteration 0: log likelihood = -20819.701
Iteration 1: log likelihood = -19799.402
Iteration 2: log likelihood = -19778.024
Iteration 3: log likelihood = -19777.969
Iteration 4: log likelihood = -19777.969

```

```

Multinomial logistic regression      Number of obs   =      20142
                                      LR chi2(36)       =      2083.47
                                      Prob > chi2       =      0.0000
Log likelihood = -19777.969          Pseudo R2       =      0.0500

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          |      (base outcome)
-----+-----
2
    tenure |    1.003085    .0028382      1.09   0.276    .9975372    1.008663
    exper  |    1.000467    .0037256      0.13   0.900    .9931917    1.007796
    male   |    1.447696    .0641653      8.35   0.000    1.327242    1.579082
    age    |    .9763086    .0031168     -7.51   0.000    .9702189    .9824366
pyperstudw~e |    .9723939    .0045356     -6.00   0.000    .9635448    .9813243
pyperstudb~k |    .9736411    .0047496     -5.48   0.000    .9643765    .9829948
pyperstudl~o |    .9739215    .0045426     -5.67   0.000    .9650586    .9828657
pyperstude~s |    .9934524    .0011368     -5.74   0.000    .9912267    .995683
pyperstudlep |    1.007398    .0016725      4.44   0.000    1.004125    1.010681
pyperstudm~e |    .9994604    .0016771     -0.32   0.748    .9961786    1.002753
    Asian  |    .6489475    .2959899     -0.95   0.343    .2654423    1.586533
    Black  |    1.397352    .1070644      4.37   0.000    1.202506    1.62377
    Latino |    1.370672    .1019282      4.24   0.000    1.184773    1.58574
    Other  |    .6849556    .221774     -1.17   0.243    .3631295    1.292002
    AsianMale |    2.221462    1.352866      1.31   0.190    .6733781    7.328565
    BlackMale |    1.413996    .1573847      3.11   0.002    1.136855    1.758697
    LatinoMale |    1.323382    .1288855      2.88   0.004    1.093417    1.601713
    OtherMale |    1.16675     .5508266      0.33   0.744    .4625113    2.943289
-----
3
    tenure |    1.010309    .0025529      4.06   0.000    1.005318    1.015325
    exper  |    1.050035    .0039942     12.84   0.000    1.042236    1.057893

```

male		1.788917	.0744955	13.97	0.000	1.648708	1.941049
age		.9848809	.0033268	-4.51	0.000	.9783821	.9914229
pyperstudw~e		1.019229	.0057275	3.39	0.001	1.008065	1.030517
pyperstudb~k		.9978427	.0058741	-0.37	0.714	.9863959	1.009422
pyperstudl~o		.9942509	.0056484	-1.01	0.310	.9832417	1.005383
pyperstude~s		1.024502	.0012456	19.91	0.000	1.022063	1.026946
pyperstudlep		1.009099	.0016415	5.57	0.000	1.005887	1.012321
pyperstudm~e		.9943656	.0018884	-2.98	0.003	.9906713	.9980736
Asian		.7133274	.2847114	-0.85	0.397	.3262478	1.559661
Black		.9394072	.0798344	-0.74	0.462	.7952717	1.109666
Latino		1.428066	.1060728	4.80	0.000	1.234591	1.65186
Other		.4097696	.1588152	-2.30	0.021	.191707	.8758738
AsianMale		1.905007	1.02935	1.19	0.233	.660631	5.493311
BlackMale		1.152022	.1443401	1.13	0.259	.9011798	1.472686
LatinoMale		.9327766	.0916724	-0.71	0.479	.7693472	1.130923
OtherMale		3.085111	1.501215	2.32	0.021	1.188711	8.006918

-----

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=20142)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		16.603	2	0.000
exper		187.037	2	0.000
male		215.180	2	0.000
age		64.357	2	0.000
pyperstudw~e		66.628	2	0.000
pyperstudb~k		31.336	2	0.000
pyperstudl~o		32.089	2	0.000
pyperstude~s		576.123	2	0.000
pyperstudlep		38.402	2	0.000
pyperstudm~e		9.465	2	0.009
Asian		1.408	2	0.495
Black		22.469	2	0.000
Latino		31.551	2	0.000
Other		6.892	2	0.032
AsianMale		2.512	2	0.285
BlackMale		9.714	2	0.008
LatinoMale		11.845	2	0.003
OtherMale		5.845	2	0.054

-----

\*\*\*\* Wald tests for independent variables (N=20142)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		16.530	2	0.000
exper		177.196	2	0.000
male		213.513	2	0.000
age		63.252	2	0.000
pyperstudw~e		66.793	2	0.000
pyperstudb~k		31.930	2	0.000
pyperstudl~o		32.709	2	0.000
pyperstude~s		533.973	2	0.000
pyperstudlep		38.105	2	0.000
pyperstudm~e		9.068	2	0.011
Asian		1.358	2	0.507
Black		22.965	2	0.000
Latino		31.609	2	0.000
Other		5.989	2	0.050
AsianMale		2.452	2	0.293



BlackMale		9.693	2	0.008
LatinoMale		11.828	2	0.003
OtherMale		5.397	2	0.067

\*\*\*\* Hausman tests of IIA assumption (N=20142)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-79.802	17	---	---
3		-6.145	18	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=20142)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-4688.852	-4681.022	15.660	19	0.680	for Ho
3		-4542.981	-4531.645	22.672	19	0.252	for Ho

\*\*\*\* Wald tests for combining alternatives (N=20142)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1214.514	18	0.000
2-	1	491.694	18	0.000
3-	1	1211.782	18	0.000

\*\*\*\* LR tests for combining alternatives (N=20142)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1358.777	18	0.000
2-	1	505.360	18	0.000
3-	1	1363.451	18	0.000

by region

```
. bysort region: mlogit admin tenure exper male age pyperstudwhite pyperstudblack
pyperstudlatino pyperstuddecodis pyperstudlep pyperstudmobile Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1996==1, rrr
```

-----> region = 1

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -1376.578
Iteration 1: log likelihood = -1278.5012
Iteration 2: log likelihood = -1275.1004
Iteration 3: log likelihood = -1274.8676
Iteration 4: log likelihood = -1274.8126
Iteration 5: log likelihood = -1274.8014
Iteration 6: log likelihood = -1274.799
Iteration 7: log likelihood = -1274.7985
```

Iteration 8: log likelihood = -1274.7983  
 Iteration 9: log likelihood = -1274.7983

Multinomial logistic regression	Number of obs	=	1293
	LR chi2(34)	=	203.56
	Prob > chi2	=	0.0000
Log likelihood = -1274.7983	Pseudo R2	=	0.0739

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9990478	.0104942	-0.09	0.928	.9786898	1.019829
	exper	1.034501	.0164247	2.14	0.033	1.002805	1.067199
	male	1.371475	.3942823	1.10	0.272	.7806884	2.409339
	age	.9487964	.0131926	-3.78	0.000	.9232885	.9750091
pyperstudw~e		1.156294	.1223792	1.37	0.170	.9396785	1.422844
pyperstudb~k		1.168566	.130975	1.39	0.165	.9380996	1.455652
pyperstudl~o		1.154073	.1167158	1.42	0.157	.9465602	1.40708
pyperstude~s		.9967288	.0037794	-0.86	0.388	.9893487	1.004164
pyperstudlep		.9995772	.0036136	-0.12	0.907	.9925198	1.006685
pyperstudm~e		1.004083	.0091689	0.45	0.655	.986272	1.022215
	Asian	8.33e-07	.001111	-0.01	0.992	0	.
	Black	9.06e-07	.0012058	-0.01	0.992	0	.
	Latino	1.021713	.2384879	0.09	0.927	.6466091	1.614418
	Other	5.69e-07	.0010703	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1849618	2.46e+09	0.01	0.991	0	.
	LatinoMale	1.745824	.5685632	1.71	0.087	.9221242	3.305306
	OtherMale	1.92e+12	4.03e+15	0.01	0.989	0	.
3							
	tenure	1.005931	.0103162	0.58	0.564	.9859138	1.026355
	exper	1.11498	.0221182	5.49	0.000	1.072461	1.159184
	male	.7094113	.2498917	-0.97	0.330	.3556809	1.414932
	age	.94295	.0168546	-3.29	0.001	.9104875	.9765699
pyperstudw~e		1.010611	.1022485	0.10	0.917	.8288267	1.232266
pyperstudb~k		.9571811	.1090789	-0.38	0.701	.7655833	1.196729
pyperstudl~o		.9639072	.0935808	-0.38	0.705	.7968864	1.165934
pyperstude~s		1.016987	.0053109	3.23	0.001	1.006631	1.027449
pyperstudlep		1.017354	.003938	4.44	0.000	1.009665	1.025102
pyperstudm~e		1.010871	.0097113	1.13	0.260	.9920151	1.030085
	Asian	2.81e-07	.0003887	-0.01	0.991	0	.
	Black	2.54e-06	.003585	-0.01	0.993	0	.
	Latino	1.131013	.2701679	0.52	0.606	.7081723	1.806326
	Other	3.83e-07	.0007777	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.7917891	1421.314	-0.00	1.000	0	.
	LatinoMale	2.571278	.9979089	2.43	0.015	1.201708	5.501726
	OtherMale	1.65e+13	3.67e+16	0.01	0.989	0	.

-----> region = 2

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -812.80728  
 Iteration 1: log likelihood = -760.58661  
 Iteration 2: log likelihood = -757.04136  
 Iteration 3: log likelihood = -756.18323  
 Iteration 4: log likelihood = -756.08109  
 Iteration 5: log likelihood = -756.0626  
 Iteration 6: log likelihood = -756.05859  
 Iteration 7: log likelihood = -756.05759

```

Iteration 8:  log likelihood = -756.05739
Iteration 9:  log likelihood = -756.05735
Iteration 10: log likelihood = -756.05734

```

```

Multinomial logistic regression      Number of obs   =      853
                                     LR chi2(32)        =     113.50
                                     Prob > chi2         =     0.0000
Log likelihood = -756.05734          Pseudo R2        =     0.0698

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.015866	.0177207	0.90	0.367	.9817215	1.051199
	exper	.9712062	.0210054	-1.35	0.177	.9308968	1.013261
	male	1.578083	.4310462	1.67	0.095	.9239065	2.695452
	age	.9819291	.0159527	-1.12	0.262	.9511549	1.013699
	pyperstudw~e	1.035895	.0679085	0.54	0.591	.910993	1.177923
	pyperstudb~k	1.033862	.0700794	0.49	0.623	.9052422	1.180757
	pyperstudl~o	1.032054	.0645771	0.50	0.614	.9129383	1.166711
	pyperstude~s	1.004733	.0054496	0.87	0.384	.9941081	1.015471
	pyperstudlep	.9837735	.0198095	-0.81	0.417	.9457038	1.023376
	pyperstudm~e	1.002022	.0090721	0.22	0.823	.9843978	1.019961
	Asian	(omitted)					
	Black	3.920933	2.296221	2.33	0.020	1.244225	12.35605
	Latino	.9542578	.2725833	-0.16	0.870	.5451555	1.670364
	Other	6.008311	7.505031	1.44	0.151	.5194091	69.50168
	AsianMale	(omitted)					
	BlackMale	.5652108	.5506256	-0.59	0.558	.0837479	3.814581
	LatinoMale	1.429823	.576024	0.89	0.375	.6491781	3.149203
	OtherMale	1.020056	1.808668	0.01	0.991	.0315757	32.95299
3							
	tenure	.9918381	.0129126	-0.63	0.529	.9668501	1.017472
	exper	1.09103	.0217174	4.38	0.000	1.049284	1.134437
	male	1.71906	.4318119	2.16	0.031	1.050696	2.812581
	age	.9732199	.0168129	-1.57	0.116	.9408188	1.006737
	pyperstudw~e	1.077973	.0604231	1.34	0.180	.9658188	1.20315
	pyperstudb~k	1.070275	.0628045	1.16	0.247	.9539956	1.200728
	pyperstudl~o	1.05246	.0562545	0.96	0.339	.9477824	1.1687
	pyperstude~s	1.020449	.0055013	3.75	0.000	1.009723	1.031288
	pyperstudlep	1.015759	.0148552	1.07	0.285	.987057	1.045296
	pyperstudm~e	.9947443	.0092885	-0.56	0.573	.9767047	1.013117
	Asian	(omitted)					
	Black	5.61e-07	.0004504	-0.02	0.986	0	.
	Latino	1.187462	.3029678	0.67	0.501	.7201867	1.957918
	Other	4.72e-07	.0017349	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	3884648	3.12e+09	0.02	0.985	0	.
	LatinoMale	1.20568	.4298559	0.52	0.600	.5994553	2.424975
	OtherMale	1.159944	5748.055	0.00	1.000	0	.

```

-----> region = 3

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0:  log likelihood = -330.41247
Iteration 1:  log likelihood = -294.83445
Iteration 2:  log likelihood = -292.58138
Iteration 3:  log likelihood = -292.42033
Iteration 4:  log likelihood = -292.39014
Iteration 5:  log likelihood = -292.38308

```

```

Iteration 6:  log likelihood = -292.38138
Iteration 7:  log likelihood = -292.38104
Iteration 8:  log likelihood = -292.38098
Iteration 9:  log likelihood = -292.38097

```

```

Multinomial logistic regression
Log likelihood = -292.38097

Number of obs   =      325
LR chi2(30)     =      76.06
Prob > chi2     =      0.0000
Pseudo R2      =      0.1151

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.019822	.027037	0.74	0.459	.9681838	1.074215
	exper	.9489836	.0350534	-1.42	0.156	.8827083	1.020235
	male	.947401	.3778096	-0.14	0.892	.4335979	2.070048
	age	.9886577	.0320004	-0.35	0.725	.9278861	1.053409
pyperstudw~e		.9505264	.052458	-0.92	0.358	.853076	1.059109
pyperstudb~k		.9913608	.0533896	-0.16	0.872	.8920525	1.101725
pyperstudl~o		.9707222	.0541893	-0.53	0.595	.8701172	1.082959
pyperstude~s		.9971432	.0155542	-0.18	0.854	.9671188	1.0281
pyperstudlep		.9581798	.0473983	-0.86	0.388	.8696421	1.055731
pyperstudm~e		1.028935	.0274409	1.07	0.285	.9765335	1.084149
	Asian	(omitted)					
	Black	3.725005	3.081334	1.59	0.112	.7362251	18.84703
	Latino	6.050358	5.892701	1.85	0.065	.8969324	40.81337
	Other	1.59e-06	.001986	-0.01	0.991	0	.
	AsianMale	(omitted)					
	BlackMale	.3155138	.4924884	-0.74	0.460	.0148037	6.724604
	LatinoMale	.3087165	.3747045	-0.97	0.333	.0286032	3.332002
	OtherMale	(omitted)					
3							
	tenure	.987604	.0181381	-0.68	0.497	.9526862	1.023802
	exper	1.038269	.0304594	1.28	0.200	.9802535	1.099718
	male	3.010259	.9627622	3.45	0.001	1.608298	5.634315
	age	.9998725	.0283275	-0.00	0.996	.9458649	1.056964
pyperstudw~e		1.133947	.0582045	2.45	0.014	1.025419	1.253961
pyperstudb~k		1.125952	.0574173	2.33	0.020	1.018857	1.244304
pyperstudl~o		1.091657	.0561226	1.71	0.088	.9870187	1.207388
pyperstude~s		1.041304	.0129945	3.24	0.001	1.016144	1.067087
pyperstudlep		1.084161	.0387005	2.26	0.024	1.010902	1.162729
pyperstudm~e		1.033105	.0225848	1.49	0.136	.9897743	1.078332
	Asian	(omitted)					
	Black	5.683454	4.791849	2.06	0.039	1.088792	29.66742
	Latino	1.447489	1.875279	0.29	0.775	.1142484	18.3392
	Other	1.237779	1.830215	0.14	0.885	.068239	22.45192
	AsianMale	(omitted)					
	BlackMale	1.49e-07	.000114	-0.02	0.984	0	.
	LatinoMale	1.175295	1.684644	0.11	0.910	.0708029	19.50933
	OtherMale	(omitted)					

-----> region = 4

```

Iteration 0:  log likelihood = -4176.9289
Iteration 1:  log likelihood = -3917.3519
Iteration 2:  log likelihood = -3906.0806
Iteration 3:  log likelihood = -3906.0498
Iteration 4:  log likelihood = -3906.0498

```

```

Multinomial logistic regression
Number of obs   =      4120
LR chi2(36)     =      541.76

```

Log likelihood = -3906.0498      Prob > chi2 = 0.0000  
Pseudo R2 = 0.0649

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9967499	.0061914	-0.52	0.600	.9846886	1.008959
	exper	1.045606	.0097182	4.80	0.000	1.026731	1.064828
	male	1.29803	.1244463	2.72	0.007	1.075666	1.566362
	age	.9470532	.0073963	-6.97	0.000	.9326671	.9616611
	pyperstudw~e	.9811366	.0068464	-2.73	0.006	.9678093	.9946474
	pyperstudb~k	.97631	.0073323	-3.19	0.001	.9620442	.9907873
	pyperstudl~o	.9815797	.0068747	-2.65	0.008	.9681976	.9951468
	pyperstude~s	1.001422	.0027563	0.52	0.606	.9960345	1.006839
	pyperstudlep	.9996658	.0048827	-0.07	0.945	.9901416	1.009282
	pyperstudm~e	.9991592	.0032868	-0.26	0.798	.9927379	1.005622
	Asian	.7813166	.6446338	-0.30	0.765	.1550722	3.93659
	Black	.8649417	.1096401	-1.14	0.252	.6746654	1.108882
	Latino	1.122785	.2191232	0.59	0.553	.7659062	1.645953
	Other	.733028	.5026319	-0.45	0.651	.1911868	2.810498
	AsianMale	.6126515	.703686	-0.43	0.670	.0644958	5.819638
	BlackMale	1.590014	.2951526	2.50	0.012	1.105079	2.28775
	LatinoMale	1.485974	.460152	1.28	0.201	.8098886	2.726449
	OtherMale	3.872889	4.780236	1.10	0.273	.3446771	43.51686
3							
	tenure	1.038768	.0071358	5.54	0.000	1.024876	1.052849
	exper	1.093163	.0125085	7.78	0.000	1.06892	1.117957
	male	1.355768	.1471562	2.80	0.005	1.095961	1.677164
	age	.956618	.0092333	-4.59	0.000	.9386911	.9748873
	pyperstudw~e	1.010267	.0094077	1.10	0.273	.9919958	1.028876
	pyperstudb~k	.9935064	.0100817	-0.64	0.521	.9739418	1.013464
	pyperstudl~o	.9858525	.0093235	-1.51	0.132	.9677471	1.004297
	pyperstude~s	1.026302	.003464	7.69	0.000	1.019535	1.033114
	pyperstudlep	1.006718	.0056131	1.20	0.230	.9957761	1.017779
	pyperstudm~e	.9924465	.004519	-1.67	0.096	.9836289	1.001343
	Asian	.8353587	.7160146	-0.21	0.834	.1556966	4.48195
	Black	.4267465	.070245	-5.17	0.000	.3090703	.5892272
	Latino	2.163827	.4237639	3.94	0.000	1.474087	3.176302
	Other	.85819	.6947223	-0.19	0.850	.1755975	4.194194
	AsianMale	.261716	.3640736	-0.96	0.335	.0171286	3.998879
	BlackMale	1.700221	.4009486	2.25	0.024	1.070962	2.699212
	LatinoMale	.9024606	.3057419	-0.30	0.762	.4645712	1.75309
	OtherMale	4.33097	6.536739	0.97	0.331	.2248359	83.42663

-----> region = 5

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -533.2813  
Iteration 1: log likelihood = -482.45775  
Iteration 2: log likelihood = -480.99412  
Iteration 3: log likelihood = -480.88836  
Iteration 4: log likelihood = -480.86394  
Iteration 5: log likelihood = -480.85904  
Iteration 6: log likelihood = -480.85825  
Iteration 7: log likelihood = -480.85805  
Iteration 8: log likelihood = -480.85801  
Iteration 9: log likelihood = -480.85801

Multinomial logistic regression      Number of obs = 549  
LR chi2(32) = 104.85

Log likelihood = -480.85801

Prob > chi2 = 0.0000  
Pseudo R2 = 0.0983

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.049497	.020363	2.49	0.013	1.010336	1.090177
	exper	.9852337	.0285791	-0.51	0.608	.9307822	1.042871
	male	3.201355	.9301463	4.00	0.000	1.81142	5.657812
	age	.9824908	.0239136	-0.73	0.468	.9367215	1.030497
pyperstudw~e		.9687161	.039077	-0.79	0.431	.895076	1.048415
pyperstudb~k		.9658761	.0405825	-0.83	0.409	.8895229	1.048783
pyperstudl~o		.9940104	.045908	-0.13	0.897	.9079846	1.088187
pyperstude~s		1.003518	.0091204	0.39	0.699	.9858007	1.021554
pyperstudlep		.9596376	.0487722	-0.81	0.418	.8686527	1.060152
pyperstudm~e		.9788396	.0243638	-0.86	0.390	.9322335	1.027776
	Asian	1.297686	7571.443	0.00	1.000	0	.
	Black	1.713447	.7139952	1.29	0.196	.7571364	3.877638
	Latino	2.93e-06	.0060095	-0.01	0.995	0	.
	Other	5.46e-07	.0011202	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.7940634	.4559335	-0.40	0.688	.2576999	2.446787
	LatinoMale	.2030472	466.0869	-0.00	0.999	0	.
	OtherMale	(omitted)					
3							
	tenure	1.02873	.0166721	1.75	0.081	.9965663	1.061931
	exper	1.041972	.0278175	1.54	0.124	.9888522	1.097945
	male	4.06749	1.129188	5.05	0.000	2.360598	7.008596
	age	.9840489	.0236738	-0.67	0.504	.938726	1.03156
pyperstudw~e		1.006676	.0387698	0.17	0.863	.9334854	1.085604
pyperstudb~k		.9751597	.0390599	-0.63	0.530	.9015318	1.054801
pyperstudl~o		1.011472	.0449737	0.26	0.798	.9270568	1.103574
pyperstude~s		1.038316	.0088037	4.43	0.000	1.021204	1.055716
pyperstudlep		.9800021	.0431699	-0.46	0.647	.8989404	1.068373
pyperstudm~e		1.019621	.0186207	1.06	0.287	.9837704	1.056778
	Asian	3.33e+07	9.63e+10	0.01	0.995	0	.
	Black	1.431633	.6341794	0.81	0.418	.6008547	3.411094
	Latino	4.90e-06	.0089976	-0.01	0.995	0	.
	Other	8.20e-07	.0015037	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1.024685	.6000452	0.04	0.967	.325189	3.228831
	LatinoMale	153301.3	2.81e+08	0.01	0.995	0	.
	OtherMale	(omitted)					

-----> region = 6

Iteration 0: log likelihood = -685.67047  
Iteration 1: log likelihood = -638.19197  
Iteration 2: log likelihood = -636.48397  
Iteration 3: log likelihood = -636.2383  
Iteration 4: log likelihood = -636.19356  
Iteration 5: log likelihood = -636.1837  
Iteration 6: log likelihood = -636.18141  
Iteration 7: log likelihood = -636.18088  
Iteration 8: log likelihood = -636.18077  
Iteration 9: log likelihood = -636.18075

Multinomial logistic regression

Number of obs = 658  
LR chi2(36) = 98.98  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0722

Log likelihood = -636.18075

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9769461	.0154599	-1.47	0.141	.9471103	1.007722
	exper	1.013005	.0247446	0.53	0.597	.9656496	1.062684
	male	1.710881	.3885054	2.36	0.018	1.096301	2.669992
	age	.9802444	.0198847	-0.98	0.325	.9420357	1.020003
pyperstudw~e		1.007569	.0540368	0.14	0.888	.9070348	1.119246
pyperstudb~k		1.005605	.0543792	0.10	0.918	.9044779	1.11804
pyperstudl~o		.9973763	.053362	-0.05	0.961	.8980855	1.107644
pyperstude~s		.9888677	.0074065	-1.49	0.135	.9744573	1.003491
pyperstudlep		1.067175	.0277726	2.50	0.012	1.014107	1.12302
pyperstudm~e		.9943764	.0130406	-0.43	0.667	.9691431	1.020267
	Asian	4.88e-07	.0013212	-0.01	0.996	0	.
	Black	.9426513	.6578197	-0.08	0.933	.2400759	3.701293
	Latino	.4349174	.5068658	-0.71	0.475	.0442987	4.269948
	Other	4.54e-07	.0012301	-0.01	0.996	0	.
	AsianMale	5.34e+13	2.51e+17	0.01	0.995	0	.
	BlackMale	4.40635	3.866661	1.69	0.091	.7891011	24.60512
	LatinoMale	5.416412	7.706103	1.19	0.235	.3331809	88.05284
	OtherMale	1.50226	4660.708	0.00	1.000	0	.
3							
	tenure	.9869457	.0137325	-0.94	0.345	.9603942	1.014231
	exper	1.095904	.0273003	3.68	0.000	1.043682	1.150739
	male	2.370904	.4951877	4.13	0.000	1.57446	3.570229
	age	.9435597	.0205367	-2.67	0.008	.9041549	.9846818
pyperstudw~e		1.049028	.0565085	0.89	0.374	.9439198	1.165841
pyperstudb~k		1.051301	.0566457	0.93	0.353	.945939	1.168399
pyperstudl~o		1.03414	.0554702	0.63	0.531	.9309406	1.148781
pyperstude~s		1.015123	.0064333	2.37	0.018	1.002592	1.027811
pyperstudlep		1.040146	.025581	1.60	0.109	.9911974	1.091512
pyperstudm~e		1.014048	.0111721	1.27	0.205	.9923854	1.036183
	Asian	4.73e-07	.0011316	-0.01	0.995	0	.
	Black	.7889455	.5012735	-0.37	0.709	.2271004	2.740793
	Latino	6.04e-07	.0005477	-0.02	0.987	0	.
	Other	5.26e-07	.0012591	-0.01	0.995	0	.
	AsianMale	1027022	6.96e+09	0.00	0.998	0	.
	BlackMale	.8079445	.7460844	-0.23	0.817	.1322367	4.936407
	LatinoMale	2298205	2.08e+09	0.02	0.987	0	.
	OtherMale	3910456	9.36e+09	0.01	0.995	0	.

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1002.4043  
Iteration 1: log likelihood = -914.85988  
Iteration 2: log likelihood = -911.03035  
Iteration 3: log likelihood = -910.93823  
Iteration 4: log likelihood = -910.92028  
Iteration 5: log likelihood = -910.91609  
Iteration 6: log likelihood = -910.91512  
Iteration 7: log likelihood = -910.91492  
Iteration 8: log likelihood = -910.91488

Multinomial logistic regression

Number of obs = 1045  
LR chi2(34) = 182.98  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0913

Log likelihood = -910.91488

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.002918	.0151601	0.19	0.847	.9736408	1.033076
	exper	.9951092	.0198314	-0.25	0.806	.9569898	1.034747
	male	2.849179	.672573	4.44	0.000	1.793851	4.52536
	age	.9774504	.0174869	-1.27	0.202	.9437706	1.012332
	pyperstudw~e	.9766055	.1581978	-0.15	0.884	.7109437	1.341538
	pyperstudb~k	.9879948	.1610578	-0.07	0.941	.717788	1.359919
	pyperstudl~o	.9837691	.1608837	-0.10	0.920	.7139854	1.355492
	pyperstude~s	.9779648	.008761	-2.49	0.013	.9609434	.9952876
	pyperstudlep	1.027898	.0299044	0.95	0.344	.9709265	1.088213
	pyperstudm~e	.9573908	.0198188	-2.10	0.035	.9193241	.9970338
	Asian	1.48e+08	6.10e+11	0.00	0.996	0	.
	Black	1.760283	.6982807	1.43	0.154	.8089507	3.830388
	Latino	2.786347	2.665725	1.07	0.284	.4272503	18.17138
	Other	4.44e-06	.003801	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	1.584601	.8415091	0.87	0.386	.5596104	4.48698
	LatinoMale	1.456973	2.522084	0.22	0.828	.0489762	43.34294
	OtherMale	.905913	7.76e+08	0.02	0.987	0	.
3							
	tenure	.9909149	.0105799	-0.85	0.393	.970394	1.01187
	exper	1.058345	.0167164	3.59	0.000	1.026083	1.091621
	male	2.315094	.3924988	4.95	0.000	1.660568	3.227606
	age	.9790523	.0144039	-1.44	0.150	.9512243	1.007694
	pyperstudw~e	.9778955	.118526	-0.18	0.854	.7711208	1.240116
	pyperstudb~k	.9508966	.1161547	-0.41	0.680	.7484394	1.20812
	pyperstudl~o	.955831	.1170452	-0.37	0.712	.7518794	1.215106
	pyperstude~s	1.04188	.0064989	6.58	0.000	1.02922	1.054696
	pyperstudlep	1.002011	.0227611	0.09	0.930	.9583789	1.04763
	pyperstudm~e	.9997963	.0098329	-0.02	0.983	.9807088	1.019255
	Asian	3.062879	21460.03	0.00	1.000	0	.
	Black	.6645246	.23435	-1.16	0.247	.3329108	1.32646
	Latino	.3751504	.43277	-0.85	0.395	.0391082	3.598682
	Other	5.15e-06	.0030378	-0.02	0.984	0	.
	AsianMale	(omitted)					
	BlackMale	2.722419	1.326068	2.06	0.040	1.047953	7.072423
	LatinoMale	2.113593	3.901328	0.41	0.685	.0567349	78.73946
	OtherMale	1293449	7.64e+08	0.02	0.981	0	.

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -311.1946  
Iteration 1: log likelihood = -277.16961  
Iteration 2: log likelihood = -275.09098  
Iteration 3: log likelihood = -274.90562  
Iteration 4: log likelihood = -274.8636  
Iteration 5: log likelihood = -274.85541  
Iteration 6: log likelihood = -274.85367  
Iteration 7: log likelihood = -274.85326  
Iteration 8: log likelihood = -274.85316  
Iteration 9: log likelihood = -274.85314

Multinomial logistic regression	Number of obs	=	319
	LR chi2(34)	=	72.68
	Prob > chi2	=	0.0001
Log likelihood = -274.85314	Pseudo R2	=	0.1168

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
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1		(base outcome)					
2							
	tenure	.9930851	.0252217	-0.27	0.785	.9448618	1.04377
	exper	1.086138	.0497845	1.80	0.071	.9928171	1.188231
	male	2.515165	1.045772	2.22	0.027	1.113393	5.681784
	age	.9355633	.039245	-1.59	0.112	.8617217	1.015733
	pyperstudw~e	.8072262	.0902804	-1.91	0.056	.6483308	1.005064
	pyperstudb~k	.8371085	.0925979	-1.61	0.108	.6739457	1.039773
	pyperstudl~o	.8798913	.1015516	-1.11	0.268	.7017599	1.103239
	pyperstude~s	.960157	.0156775	-2.49	0.013	.9299162	.9913813
	pyperstudlep	.8835022	.0864887	-1.27	0.206	.7292577	1.070371
	pyperstudm~e	.9612512	.0340796	-1.11	0.265	.8967243	1.030421
	Asian	3.36e-07	.0013579	-0.00	0.997	0	.
	Black	3.687916	3.076765	1.56	0.118	.7188501	18.92011
	Latino	1.37e-06	.0036935	-0.01	0.996	0	.
	Other	8.40e-07	.0012991	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	.9579048	1.107769	-0.04	0.970	.0993003	9.24047
	LatinoMale	.8968664	6.57e+10	0.00	0.998	0	.
	OtherMale	783838.1	5.48e+09	0.00	0.998	0	.
3							
	tenure	.9934182	.0190904	-0.34	0.731	.9566976	1.031548
	exper	1.06248	.0299774	2.15	0.032	1.00532	1.122889
	male	3.056625	.9223699	3.70	0.000	1.691932	5.522065
	age	.9698344	.0242452	-1.23	0.220	.9234602	1.018537
	pyperstudw~e	.9570043	.1010253	-0.42	0.677	.7781396	1.176983
	pyperstudb~k	.9328783	.0986961	-0.66	0.511	.7581761	1.147836
	pyperstudl~o	1.002048	.1113106	0.02	0.985	.8059991	1.245784
	pyperstude~s	1.029007	.0108131	2.72	0.007	1.00803	1.05042
	pyperstudlep	.9415796	.0525157	-1.08	0.280	.8440771	1.050345
	pyperstudm~e	1.016703	.0142754	1.18	0.238	.9891057	1.045071
	Asian	2.30e-07	.0006148	-0.01	0.995	0	.
	Black	2.807485	1.926452	1.50	0.132	.7315367	10.77454
	Latino	4.62e-07	.0008352	-0.01	0.994	0	.
	Other	1.025509	1.249067	0.02	0.984	.0942276	11.16094
	AsianMale	(omitted)					
	BlackMale	.8701024	.8775458	-0.14	0.890	.1205278	6.281359
	LatinoMale	1.16e+14	4.19e+17	0.01	0.993	0	.
	OtherMale	1.35e+07	4.20e+10	0.01	0.996	0	.

-----> region = 9

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -214.48902  
Iteration 1: log likelihood = -191.07223  
Iteration 2: log likelihood = -189.32473  
Iteration 3: log likelihood = -189.12684  
Iteration 4: log likelihood = -189.08202  
Iteration 5: log likelihood = -189.07321  
Iteration 6: log likelihood = -189.07166  
Iteration 7: log likelihood = -189.07131  
Iteration 8: log likelihood = -189.07122  
Iteration 9: log likelihood = -189.07121

Multinomial logistic regression	Number of obs	=	219
	LR chi2(30)	=	50.84
	Prob > chi2	=	0.0101
Log likelihood = -189.07121	Pseudo R2	=	0.1185

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.072287	.0482337	1.55	0.121	.9817979	1.171115
	exper	.9280686	.057253	-1.21	0.226	.8223734	1.047348
	male	2.347062	1.218964	1.64	0.100	.8480992	6.495346
	age	.9778539	.040655	-0.54	0.590	.9013316	1.060873
pyperstudw~e		.8304926	.1122358	-1.37	0.169	.6372374	1.082356
pyperstudb~k		.8508792	.1218193	-1.13	0.259	.6426913	1.126506
pyperstudl~o		.8129578	.1123986	-1.50	0.134	.619986	1.065992
pyperstude~s		1.005996	.0192914	0.31	0.755	.9688871	1.044526
pyperstudlep		.9750025	.1590465	-0.16	0.877	.708197	1.342324
pyperstudm~e		.9559936	.0469306	-0.92	0.359	.8682979	1.052546
	Asian	(omitted)					
	Black	16.7206	23.12842	2.04	0.042	1.111337	251.5697
	Latino	13.9352	43438.26	0.00	0.999	0	.
	Other	1.72e-07	.0004419	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	2.56e-08	.0000362	-0.01	0.990	0	.
	LatinoMale	8.37e-08	.0002827	-0.00	0.996	0	.
	OtherMale	(omitted)					
3							
	tenure	.9980618	.0230603	-0.08	0.933	.9538725	1.044298
	exper	1.031512	.0368658	0.87	0.385	.9617293	1.106359
	male	3.117158	1.066702	3.32	0.001	1.59395	6.095971
	age	1.004849	.0296688	0.16	0.870	.9483495	1.064714
pyperstudw~e		1.097352	.1346235	0.76	0.449	.8628205	1.395635
pyperstudb~k		1.113042	.1480931	0.80	0.421	.8575453	1.444661
pyperstudl~o		1.073597	.1317323	0.58	0.563	.8441063	1.36548
pyperstude~s		1.035093	.013226	2.70	0.007	1.009492	1.061343
pyperstudlep		1.051852	.0950831	0.56	0.576	.8810681	1.25574
pyperstudm~e		.9767245	.0225304	-1.02	0.307	.9335491	1.021897
	Asian	(omitted)					
	Black	4.24998	5.291561	1.16	0.245	.3703169	48.77532
	Latino	1374322	1.96e+09	0.01	0.992	0	.
	Other	1.26e-06	.001783	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	1.15e-07	.0000937	-0.02	0.984	0	.
	LatinoMale	4.64e-07	.0006615	-0.01	0.992	0	.
	OtherMale	(omitted)					

-----> region = 10

Iteration 0: log likelihood = -2758.2734  
Iteration 1: log likelihood = -2571.8378  
Iteration 2: log likelihood = -2567.7383  
Iteration 3: log likelihood = -2567.7093  
Iteration 4: log likelihood = -2567.7093

Multinomial logistic regression	Number of obs	=	2659
	LR chi2(36)	=	381.13
	Prob > chi2	=	0.0000
Log likelihood = -2567.7093	Pseudo R2	=	0.0691

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.00939	.0078779	1.20	0.231	.9940671	1.024949
	exper	1.03011	.0118728	2.57	0.010	1.0071	1.053645

male		1.768534	.2146266	4.70	0.000	1.394161	2.243437
age		.9591049	.0093993	-4.26	0.000	.9408584	.9777053
pyperstudw~e		.994958	.0128224	-0.39	0.695	.9701413	1.02041
pyperstudb~k		.997862	.0130043	-0.16	0.870	.9726968	1.023678
pyperstudl~o		.9983924	.0130127	-0.12	0.902	.9732109	1.024225
pyperstude~s		.9928941	.003901	-1.82	0.070	.9852776	1.000569
pyperstudlep		1.002472	.0073593	0.34	0.737	.9881513	1.017
pyperstudm~e		.9926472	.0059299	-1.24	0.217	.9810926	1.004338
Asian		.5411159	.5853555	-0.57	0.570	.0649378	4.509028
Black		2.065103	.3816079	3.92	0.000	1.437637	2.966431
Latino		3.695645	.9820494	4.92	0.000	2.195326	6.221302
Other		.6180363	.4773536	-0.62	0.533	.1360098	2.808392
AsianMale		2.914423	3.605687	0.86	0.387	.2579034	32.93428
BlackMale		1.258428	.3348541	0.86	0.388	.7470209	2.119942
LatinoMale		.5945453	.226825	-1.36	0.173	.2814771	1.255819
OtherMale		.2927495	.381338	-0.94	0.346	.0227884	3.760785
-----							
3							
tenure		1.021479	.0074679	2.91	0.004	1.006947	1.036222
exper		1.095762	.0135151	7.41	0.000	1.069591	1.122574
male		1.507014	.1769355	3.49	0.000	1.197235	1.896947
age		.9505108	.0100941	-4.78	0.000	.9309312	.9705022
pyperstudw~e		1.040191	.0140377	2.92	0.004	1.013039	1.068072
pyperstudb~k		1.01261	.0141393	0.90	0.369	.9852728	1.040705
pyperstudl~o		1.00264	.0141144	0.19	0.851	.9753545	1.030689
pyperstude~s		1.025293	.0040066	6.39	0.000	1.01747	1.033176
pyperstudlep		1.00724	.0077296	0.94	0.347	.9922035	1.022504
pyperstudm~e		.9917855	.0062654	-1.31	0.192	.9795813	1.004142
Asian		.5560537	.401728	-0.81	0.417	.1349446	2.291279
Black		1.13758	.2345399	0.63	0.532	.7594265	1.704032
Latino		2.931838	.8328629	3.79	0.000	1.680098	5.116175
Other		.1682925	.179549	-1.67	0.095	.0207936	1.362074
AsianMale		5.534601	4.914622	1.93	0.054	.9710355	31.5455
BlackMale		1.645842	.4980258	1.65	0.100	.9095315	2.978233
LatinoMale		.9134465	.3761001	-0.22	0.826	.4075788	2.047173
OtherMale		7.502227	9.121192	1.66	0.097	.6923109	81.29788
-----							

-----> region = 11

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1987.0581  
Iteration 1: log likelihood = -1824.8319  
Iteration 2: log likelihood = -1822.1889  
Iteration 3: log likelihood = -1821.9681  
Iteration 4: log likelihood = -1821.9192  
Iteration 5: log likelihood = -1821.9089  
Iteration 6: log likelihood = -1821.9071  
Iteration 7: log likelihood = -1821.9069  
Iteration 8: log likelihood = -1821.9069  
Iteration 9: log likelihood = -1821.9069

Multinomial logistic regression	Number of obs	=	1850
	LR chi2(34)	=	330.30
	Prob > chi2	=	0.0000
Log likelihood = -1821.9069	Pseudo R2	=	0.0831

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9753914	.0095764	-2.54	0.011	.9568014 .9943426
exper		.9528169	.0097327	-4.73	0.000	.9339309 .9720848
male		1.128761	.1476015	0.93	0.354	.8735654 1.458508

age		1.006101	.0088089	0.69	0.487	.9889834	1.023515
pyperstudw~e		.9429411	.0176265	-3.14	0.002	.9090191	.978129
pyperstudb~k		.9320947	.0181481	-3.61	0.000	.8971953	.9683517
pyperstudl~o		.926932	.0171695	-4.10	0.000	.8938839	.961202
pyperstude~s		.9982685	.004193	-0.41	0.680	.9900842	1.00652
pyperstudlep		1.003204	.0132306	0.24	0.808	.9776048	1.029474
pyperstudm~e		.9938326	.0051806	-1.19	0.235	.9837306	1.004038
Asian		.2503123	.2843002	-1.22	0.223	.0270211	2.318791
Black		2.96924	.8890921	3.63	0.000	1.651074	5.33979
Latino		3.121323	1.132299	3.14	0.002	1.533045	6.355103
Other		.1533719	.1659298	-1.73	0.083	.0184013	1.278329
AsianMale		(omitted)					
BlackMale		1.132004	.4691677	0.30	0.765	.5024135	2.550556
LatinoMale		.6060017	.3261002	-0.93	0.352	.2110704	1.739884
OtherMale		3.62e-06	.0045465	-0.01	0.992	0	.

3							
tenure		.9880901	.0089772	-1.32	0.187	.9706509	1.005843
exper		.9729155	.0099042	-2.70	0.007	.953696	.9925222
male		1.7117	.2223016	4.14	0.000	1.327029	2.207876
age		1.058741	.0097394	6.21	0.000	1.039823	1.078003
pyperstudw~e		1.020078	.0208197	0.97	0.330	.9800779	1.061711
pyperstudb~k		.9878955	.0210393	-0.57	0.567	.9475079	1.030005
pyperstudl~o		.9752944	.0198786	-1.23	0.220	.9371011	1.015044
pyperstude~s		1.032864	.0044697	7.47	0.000	1.024141	1.041662
pyperstudlep		1.018745	.0139242	1.36	0.174	.9918165	1.046405
pyperstudm~e		.9831693	.0067611	-2.47	0.014	.9700066	.9965106
Asian		.5551733	.6224433	-0.52	0.600	.0616719	4.997694
Black		3.626295	1.163675	4.01	0.000	1.933362	6.80163
Latino		1.928299	.7949731	1.59	0.111	.8595134	4.326097
Other		4.17e-07	.0003399	-0.02	0.986	0	.
AsianMale		(omitted)					
BlackMale		.3119982	.1549119	-2.35	0.019	.117901	.8256324
LatinoMale		.963012	.5533627	-0.07	0.948	.3122605	2.969931
OtherMale		666313.7	5.43e+08	0.02	0.987	0	.

-----> region = 12

Iteration 0: log likelihood = -763.3785  
Iteration 1: log likelihood = -693.73895  
Iteration 2: log likelihood = -690.25553  
Iteration 3: log likelihood = -689.95535  
Iteration 4: log likelihood = -689.90916  
Iteration 5: log likelihood = -689.90623  
Iteration 6: log likelihood = -689.90559  
Iteration 7: log likelihood = -689.90544  
Iteration 8: log likelihood = -689.90541

Multinomial logistic regression	Number of obs	=	740
	LR chi2(36)	=	146.95
	Prob > chi2	=	0.0000
Log likelihood = -689.90541	Pseudo R2	=	0.0962

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9955904	.0167873	-0.26	0.793	.9632256 1.029043
exper		1.004688	.0233406	0.20	0.840	.9599667 1.051492
male		1.143506	.2776885	0.55	0.581	.7104494 1.840534
age		.943782	.0189493	-2.88	0.004	.9073633 .9816624
pyperstudw~e		.7534871	.043999	-4.85	0.000	.6720026 .8448521
pyperstudb~k		.7527914	.0482645	-4.43	0.000	.6638969 .8535887

pyperstudl~o		.7722882	.0464266	-4.30	0.000	.68645	.8688602
pyperstude~s		.9990331	.0085636	-0.11	0.910	.982389	1.015959
pyperstudlep		.9940849	.0277562	-0.21	0.832	.9411455	1.050002
pyperstudm~e		.9962063	.0083835	-0.45	0.652	.9799096	1.012774
Asian		.0000128	.0137248	-0.01	0.992	0	.
Black		1.912894	.858083	1.45	0.148	.7940727	4.608096
Latino		13.42091	14.93914	2.33	0.020	1.514568	118.9256
Other		2.216242	2.841506	0.62	0.535	.1795841	27.35059
AsianMale		1.57e+11	1.86e+14	0.02	0.983	0	.
BlackMale		1.932936	1.29392	0.98	0.325	.5204966	7.178228
LatinoMale		.1267417	.1799577	-1.45	0.146	.00784	2.048904
OtherMale		9.06e-07	.0009724	-0.01	0.990	0	.

3							
tenure		1.017681	.0130193	1.37	0.171	.9924806	1.043521
exper		1.032922	.0187645	1.78	0.075	.9967915	1.070363
male		2.278311	.445317	4.21	0.000	1.553237	3.34186
age		.9749143	.0155724	-1.59	0.112	.9448658	1.005918
pyperstudw~e		.9102605	.0518982	-1.65	0.099	.8140194	1.01788
pyperstudb~k		.888032	.0536486	-1.97	0.049	.7888692	.9996598
pyperstudl~o		.8933431	.0523447	-1.92	0.054	.7964212	1.00206
pyperstude~s		1.027143	.0070704	3.89	0.000	1.013379	1.041095
pyperstudlep		1.00883	.0245679	0.36	0.718	.9618086	1.05815
pyperstudm~e		.9908785	.0083714	-1.08	0.278	.974606	1.007423
Asian		5.95e-06	.0049449	-0.01	0.988	0	.
Black		2.180131	.9117317	1.86	0.062	.9605274	4.948293
Latino		6.197845	7.768439	1.46	0.146	.5312958	72.30112
Other		2.144257	2.697887	0.61	0.544	.1820966	25.24945
AsianMale		4.51e+10	4.37e+13	0.03	0.980	0	.
BlackMale		.6260038	.397892	-0.74	0.461	.1801142	2.175735
LatinoMale		.182224	.273858	-1.13	0.257	.00958	3.46614
OtherMale		1.35e-06	.0011202	-0.02	0.987	0	.

-----> region = 13

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1149.566  
Iteration 1: log likelihood = -1083.2937  
Iteration 2: log likelihood = -1081.1058  
Iteration 3: log likelihood = -1080.8808  
Iteration 4: log likelihood = -1080.836  
Iteration 5: log likelihood = -1080.8264  
Iteration 6: log likelihood = -1080.8243  
Iteration 7: log likelihood = -1080.8238  
Iteration 8: log likelihood = -1080.8237  
Iteration 9: log likelihood = -1080.8237

Multinomial logistic regression	Number of obs	=	1070
	LR chi2(32)	=	137.48
	Prob > chi2	=	0.0000
Log likelihood = -1080.8237	Pseudo R2	=	0.0598

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.032077	.0125496	2.60	0.009	1.007771 1.056969
exper		.993256	.0166635	-0.40	0.687	.9611273 1.026459
male		1.210204	.2184239	1.06	0.290	.8496289 1.723804
age		.9755579	.01434	-1.68	0.092	.9478531 1.004072
pyperstudw~e		.9892583	.0340442	-0.31	0.754	.9247334 1.058285
pyperstudb~k		1.013835	.0365099	0.38	0.703	.9447442 1.087979

pyperstudl~o		1.009025	.0336437	0.27	0.788	.9451932	1.077168
pyperstude~s		.9909533	.0066148	-1.36	0.173	.9780731	1.004003
pyperstudlep		.9930139	.0152366	-0.46	0.648	.9635952	1.023331
pyperstudm~e		.9854191	.0075464	-1.92	0.055	.9707388	1.000321
Asian		(omitted)					
Black		1.607124	.5803617	1.31	0.189	.7918901	3.261623
Latino		1.04364	.3252821	0.14	0.891	.5665625	1.922443
Other		6.29e-07	.0004911	-0.02	0.985	0	.
AsianMale		(omitted)					
BlackMale		5.018131	4.345145	1.86	0.062	.9193817	27.38976
LatinoMale		2.243817	1.056578	1.72	0.086	.8915999	5.646832
OtherMale		646502.9	5.04e+08	0.02	0.986	0	.

3							
tenure		1.014082	.0115092	1.23	0.218	.9917739	1.036893
exper		1.079307	.0189412	4.35	0.000	1.042814	1.117077
male		1.351395	.2316249	1.76	0.079	.9658044	1.890931
age		.9612018	.0153096	-2.48	0.013	.931659	.9916815
pyperstudw~e		.9931704	.0329062	-0.21	0.836	.9307249	1.059806
pyperstudb~k		.9800416	.034476	-0.57	0.567	.9147468	1.049997
pyperstudl~o		.9799285	.0316856	-0.63	0.531	.9197528	1.044041
pyperstude~s		1.021813	.006906	3.19	0.001	1.008367	1.035439
pyperstudlep		.9901596	.016157	-0.61	0.544	.9589934	1.022339
pyperstudm~e		.9856173	.0077585	-1.84	0.066	.9705277	1.000942
Asian		(omitted)					
Black		1.197288	.4901564	0.44	0.660	.5366935	2.670983
Latino		1.054368	.3329873	0.17	0.867	.5677649	1.958016
Other		7.91e-07	.0006639	-0.02	0.987	0	.
AsianMale		(omitted)					
BlackMale		8.929475	7.834641	2.50	0.013	1.599519	49.84969
LatinoMale		1.249987	.6356363	0.44	0.661	.4613781	3.38652
OtherMale		2.110636	2178.939	0.00	0.999	0	.

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -268.3608  
Iteration 1: log likelihood = -238.01883  
Iteration 2: log likelihood = -234.99729  
Iteration 3: log likelihood = -233.75169  
Iteration 4: log likelihood = -233.5936  
Iteration 5: log likelihood = -233.56051  
Iteration 6: log likelihood = -233.55552  
Iteration 7: log likelihood = -233.55492  
Iteration 8: log likelihood = -233.5548  
Iteration 9: log likelihood = -233.55478  
Iteration 10: log likelihood = -233.55477

Multinomial logistic regression	Number of obs	=	284
	LR chi2(32)	=	69.61
	Prob > chi2	=	0.0001
Log likelihood = -233.55477	Pseudo R2	=	0.1297

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9981351	.0304542	-0.06	0.951	.9401957 1.059645
exper		1.044098	.0558675	0.81	0.420	.9401459 1.159544
male		1.639162	.7394024	1.10	0.273	.6771065 3.968137
age		.9379445	.0478136	-1.26	0.209	.848761 1.036499
pyperstudw~e		1.199576	.2582996	0.85	0.398	.7865786 1.82942

pyperstudb~k		1.310253	.3142129	1.13	0.260	.8188945	2.096441
pyperstudl~o		1.182022	.2468272	0.80	0.423	.7850182	1.779802
pyperstude~s		.9989124	.018063	-0.06	0.952	.9641295	1.03495
pyperstudlep		.8916741	.1209083	-0.85	0.398	.6835744	1.163125
pyperstudm~e		.9808223	.0376396	-0.50	0.614	.9097561	1.05744
Asian		(omitted)					
Black		2.84e-07	.0007756	-0.01	0.996	0	.
Latino		6.77e-07	.0020841	-0.00	0.996	0	.
Other		9.80e-07	.0032341	-0.00	0.997	0	.
AsianMale		(omitted)					
BlackMale		2091898	1.25e+10	0.00	0.998	0	.
LatinoMale		5.14e+14	3.07e+18	0.01	0.995	0	.
OtherMale		.7270303	3393.906	-0.00	1.000	0	.

3							
tenure		.9784627	.0183531	-1.16	0.246	.9431445	1.015103
exper		1.045938	.0324749	1.45	0.148	.9841865	1.111564
male		2.752858	.8392826	3.32	0.001	1.51451	5.003748
age		.9868519	.0291433	-0.45	0.654	.9313537	1.045657
pyperstudw~e		1.691539	.3127718	2.84	0.004	1.177312	2.430369
pyperstudb~k		1.653335	.3357125	2.48	0.013	1.110511	2.461494
pyperstudl~o		1.650236	.3001225	2.75	0.006	1.155423	2.356953
pyperstude~s		1.013091	.0115261	1.14	0.253	.99075	1.035935
pyperstudlep		1.07252	.0426566	1.76	0.078	.9920906	1.159471
pyperstudm~e		1.003198	.0217632	0.15	0.883	.9614371	1.046773
Asian		(omitted)					
Black		2.610441	3.345691	0.75	0.454	.2117215	32.18569
Latino		2.43e-07	.0004217	-0.01	0.993	0	.
Other		1.295155	1.947734	0.17	0.863	.0679569	24.68372
AsianMale		(omitted)					
BlackMale		6309998	1.44e+10	0.01	0.995	0	.
LatinoMale		3.84e+14	2.08e+18	0.01	0.995	0	.
OtherMale		1.421892	3.073102	0.16	0.871	.0205676	98.29905

-----> region = 15

note: Black omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -291.15973  
Iteration 1: log likelihood = -250.43163  
Iteration 2: log likelihood = -248.35463  
Iteration 3: log likelihood = -248.29836  
Iteration 4: log likelihood = -248.28637  
Iteration 5: log likelihood = -248.28347  
Iteration 6: log likelihood = -248.28286  
Iteration 7: log likelihood = -248.28273  
Iteration 8: log likelihood = -248.2827  
Iteration 9: log likelihood = -248.2827

Multinomial logistic regression	Number of obs	=	298
	LR chi2(28)	=	85.75
	Prob > chi2	=	0.0000
Log likelihood = -248.2827	Pseudo R2	=	0.1473

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.03319	.0327643	1.03	0.303	.9709278 1.099444
exper		1.085425	.062558	1.42	0.155	.9694849 1.215229
male		1.212472	.5693944	0.41	0.682	.4829866 3.043745

age		.8985206	.0459025	-2.09	0.036	.8129108	.9931461
pyperstudw~e		.9457389	.2605669	-0.20	0.840	.5511261	1.622899
pyperstudb~k		.9843325	.3036118	-0.05	0.959	.5377662	1.801732
pyperstudl~o		.9858757	.2741623	-0.05	0.959	.5716253	1.700329
pyperstude~s		.96732	.0166812	-1.93	0.054	.9351719	1.000573
pyperstudlep		.99085	.0290602	-0.31	0.754	.9354992	1.049476
pyperstudm~e		1.061931	.0347238	1.84	0.066	.9960089	1.132217
Asian		5.12e-07	.0010701	-0.01	0.994	0	.
Black		(omitted)					
Latino		.5233574	.4462176	-0.76	0.448	.0984145	2.783158
Other		5.530374	16132.67	0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		(omitted)					
LatinoMale		1.672294	1.56546	0.55	0.582	.267279	10.47118
OtherMale		(omitted)					
-----							
3							
tenure		1.013535	.0204712	0.67	0.506	.9741956	1.054462
exper		1.057723	.0349729	1.70	0.090	.9913515	1.128539
male		4.577752	1.537913	4.53	0.000	2.36967	8.843346
age		.9606676	.0286435	-1.35	0.178	.9061362	1.018481
pyperstudw~e		.8709734	.1704369	-0.71	0.480	.5935223	1.278123
pyperstudb~k		.8547981	.180684	-0.74	0.458	.5648587	1.293562
pyperstudl~o		.8395249	.165184	-0.89	0.374	.5708895	1.234568
pyperstude~s		1.041816	.0112727	3.79	0.000	1.019955	1.064146
pyperstudlep		1.003191	.0239697	0.13	0.894	.9572947	1.051289
pyperstudm~e		.9716842	.0196958	-1.42	0.156	.9338379	1.011064
Asian		5.42e-06	.0066302	-0.01	0.992	0	.
Black		(omitted)					
Latino		1.185828	.8725307	0.23	0.817	.2803555	5.015735
Other		889228.8	1.17e+09	0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		(omitted)					
LatinoMale		.877493	.6982641	-0.16	0.870	.1844595	4.174326
OtherMale		(omitted)					
-----							

-----> region = 16

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -431.37699
Iteration 1: log likelihood = -401.19283
Iteration 2: log likelihood = -398.46452
Iteration 3: log likelihood = -397.53199
Iteration 4: log likelihood = -397.38419
Iteration 5: log likelihood = -397.36069
Iteration 6: log likelihood = -397.355
Iteration 7: log likelihood = -397.35383
Iteration 8: log likelihood = -397.35358
Iteration 9: log likelihood = -397.35352
Iteration 10: log likelihood = -397.35351

```

Multinomial logistic regression	Number of obs	=	414
	LR chi2(34)	=	68.05
	Prob > chi2	=	0.0005
Log likelihood = -397.35351	Pseudo R2	=	0.0789

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.04725	.0243806	1.98	0.047	1.000539 1.096143
exper		.9795089	.0300095	-0.68	0.499	.9224225 1.040128
male		1.10881	.3484167	0.33	0.742	.5989458 2.052705



age		.9723561	.0274762	-0.99	0.321	.9199678	1.027728
pyperstudw~e		.9684863	.0314275	-0.99	0.324	.9088075	1.032084
pyperstudb~k		.9937667	.0419061	-0.15	0.882	.9149348	1.079391
pyperstudl~o		.9816257	.0306005	-0.59	0.552	.9234452	1.043472
pyperstude~s		.9805164	.0107398	-1.80	0.072	.9596911	1.001794
pyperstudlep		.9971642	.0280362	-0.10	0.920	.9437009	1.053656
pyperstudm~e		1.005573	.0134908	0.41	0.679	.9794761	1.032365
Asian		9.53e-08	.0003462	-0.00	0.996	0	.
Black		.6268645	.6130313	-0.48	0.633	.0922053	4.261785
Latino		1.027771	.9533345	0.03	0.976	.1668566	6.330667
Other		2.67e-07	.0006555	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		5.654132	9.00619	1.09	0.277	.2491889	128.2931
LatinoMale		4.155603	5.411722	1.09	0.274	.323697	53.34938
OtherMale		1.14e+07	2.79e+10	0.01	0.995	0	.
-----							
3							
tenure		1.033348	.0166187	2.04	0.041	1.001284	1.066439
exper		1.023	.0229159	1.02	0.310	.9790572	1.068915
male		1.525757	.375621	1.72	0.086	.9417397	2.471951
age		.9867211	.022165	-0.60	0.552	.944221	1.031134
pyperstudw~e		1.036617	.0344425	1.08	0.279	.9712624	1.10637
pyperstudb~k		1.040922	.0427482	0.98	0.329	.9604201	1.128171
pyperstudl~o		1.016133	.0328808	0.49	0.621	.9536891	1.082666
pyperstude~s		1.009399	.007923	1.19	0.233	.9939892	1.025048
pyperstudlep		1.04622	.0216009	2.19	0.029	1.004729	1.089426
pyperstudm~e		1.002098	.0136243	0.15	0.877	.9757475	1.02916
Asian		3.23e-07	.0008041	-0.01	0.995	0	.
Black		3.33e-07	.0002799	-0.02	0.986	0	.
Latino		.287492	.2764755	-1.30	0.195	.0436558	1.893259
Other		2.67e-07	.0004702	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		.4703347	1085.525	-0.00	1.000	0	.
LatinoMale		6.21e-07	.0009017	-0.01	0.992	0	.
OtherMale		.5652572	1362.98	-0.00	1.000	0	.
-----							

-----> region = 17

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -414.07003  
Iteration 1: log likelihood = -352.52233  
Iteration 2: log likelihood = -349.36404  
Iteration 3: log likelihood = -348.14398  
Iteration 4: log likelihood = -347.6358  
Iteration 5: log likelihood = -346.64897  
Iteration 6: log likelihood = -346.61343  
Iteration 7: log likelihood = -346.61019  
Iteration 8: log likelihood = -346.60962  
Iteration 9: log likelihood = -346.60956  
Iteration 10: log likelihood = -346.60954

Multinomial logistic regression	Number of obs	=	411
	LR chi2(32)	=	134.92
	Prob > chi2	=	0.0000
Log likelihood = -346.60954	Pseudo R2	=	0.1629

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1						
tenure		1.018149	.0184181	0.99	0.320	.9826826 1.054895
exper		.9250049	.0255899	-2.82	0.005	.8761852 .9765447
male		.3860093	.101845	-3.61	0.000	.2301527 .6474101
age		1.011621	.0252336	0.46	0.643	.9633536 1.062307

```

pyperstudw~e | 1.183638 .174765 1.14 0.254 .8862133 1.580882
pyperstudb~k | 1.214923 .1814827 1.30 0.192 .9065634 1.628169
pyperstudl~o | 1.214175 .1788778 1.32 0.188 .9096585 1.620631
pyperstude~s | .9634485 .0110271 -3.25 0.001 .9420765 .9853054
pyperstudlep | .9605415 .025348 -1.53 0.127 .9121232 1.01153
pyperstudm~e | 1.048746 .0186997 2.67 0.008 1.012728 1.086044
Asian | (omitted)
Black | .1447355 .1498194 -1.87 0.062 .0190316 1.100716
Latino | 1.170151 .8409491 0.22 0.827 .286096 4.785997
Other | 2.53e-07 .0003199 -0.01 0.990 0 .
AsianMale | (omitted)
BlackMale | 5.17e-09 .0001252 -0.00 0.999 0 .
LatinoMale | .2763922 .2597961 -1.37 0.171 .043796 1.744282
OtherMale | 2.919767 4444.857 0.00 0.999 0 .
-----
2
tenure | .9399605 .0247901 -2.35 0.019 .8926073 .9898258
exper | .9626737 .0366574 -1.00 0.318 .8934421 1.03727
male | 1.369991 .5955702 0.72 0.469 .5843571 3.211862
age | 1.020301 .0365843 0.56 0.575 .9510584 1.094584
pyperstudw~e | .9795953 .19832 -0.10 0.919 .6587493 1.45671
pyperstudb~k | 1.081637 .2204628 0.39 0.700 .7254156 1.612784
pyperstudl~o | 1.031506 .2083996 0.15 0.878 .6942239 1.532652
pyperstude~s | .9068079 .0156858 -5.66 0.000 .8765796 .9380785
pyperstudlep | 1.006866 .0438602 0.16 0.875 .9244691 1.096607
pyperstudm~e | 1.050451 .0249584 2.07 0.038 1.002655 1.100525
Asian | (omitted)
Black | .504089 .6970361 -0.50 0.620 .0335349 7.577362
Latino | 1.832913 2.359422 0.47 0.638 .1470384 22.84825
Other | 1.29e-06 .0027482 -0.01 0.995 0 .
AsianMale | (omitted)
BlackMale | 2.998823 5.386721 0.61 0.541 .0887067 101.3784
LatinoMale | .2812631 .4192892 -0.85 0.395 .0151424 5.224323
OtherMale | .331627 870.4304 -0.00 1.000 0 .
-----
3 | (base outcome)
-----

```

-----> region = 18

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -384.10273
Iteration 1: log likelihood = -338.75137
Iteration 2: log likelihood = -335.31596
Iteration 3: log likelihood = -335.10455
Iteration 4: log likelihood = -335.05631
Iteration 5: log likelihood = -335.04467
Iteration 6: log likelihood = -335.04222
Iteration 7: log likelihood = -335.04171
Iteration 8: log likelihood = -335.04159
Iteration 9: log likelihood = -335.04156

```

```

Multinomial logistic regression      Number of obs   =      391
                                     LR chi2(30)        =      98.12
                                     Prob > chi2        =      0.0000
Log likelihood = -335.04156          Pseudo R2       =      0.1277

```

```

-----
admin |          RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1      | (base outcome)
-----+-----
2      |
tenure |    1.00811    .0263283    0.31  0.757    .9578062    1.061056

```

exper		.9444131	.0374393	-1.44	0.149	.8738117	1.020719
male		1.475219	.6095054	0.94	0.347	.6564045	3.31544
age		1.00072	.034407	0.02	0.983	.9355055	1.070481
pyperstudw~e		1.048945	.2048036	0.24	0.807	.7154153	1.537969
pyperstudb~k		1.011621	.2062286	0.06	0.955	.6784101	1.508493
pyperstudl~o		1.036169	.202496	0.18	0.856	.7064513	1.519773
pyperstude~s		1.003535	.0162357	0.22	0.827	.9722127	1.035866
pyperstudlep		.9773854	.0230448	-0.97	0.332	.9332462	1.023612
pyperstudm~e		1.023618	.0136499	1.75	0.080	.9972109	1.050724
Asian		(omitted)					
Black		6.19e-06	.0062308	-0.01	0.990	0	.
Latino		4.578774	3.275141	2.13	0.033	1.126926	18.60387
Other		1.13e-06	.0017356	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		7.36e+11	8.82e+14	0.02	0.982	0	.
LatinoMale		.7363302	.6211977	-0.36	0.717	.1409175	3.847516
OtherMale		(omitted)					

3							
tenure		.9901345	.0166706	-0.59	0.556	.957994	1.023353
exper		1.140954	.0382154	3.94	0.000	1.068459	1.218368
male		1.30407	.3671899	0.94	0.346	.7509762	2.264517
age		.9143333	.0280832	-2.92	0.004	.8609152	.9710659
pyperstudw~e		.8756484	.1229919	-0.95	0.344	.6649235	1.153155
pyperstudb~k		.8183416	.1220772	-1.34	0.179	.6108795	1.09626
pyperstudl~o		.8331107	.1178302	-1.29	0.197	.6314135	1.099238
pyperstude~s		1.074531	.0138275	5.59	0.000	1.047768	1.101977
pyperstudlep		.9816139	.0141902	-1.28	0.199	.954192	1.009824
pyperstudm~e		.9940983	.0145864	-0.40	0.687	.9659167	1.023102
Asian		(omitted)					
Black		2.045585	2.607221	0.56	0.574	.1682325	24.87282
Latino		1.996645	1.144658	1.21	0.228	.6491048	6.141674
Other		2.44e-06	.0021388	-0.01	0.988	0	.
AsianMale		(omitted)					
BlackMale		662187.2	4.31e+08	0.02	0.984	0	.
LatinoMale		.4960803	.3515901	-0.99	0.323	.1236729	1.989891
OtherMale		(omitted)					

-----> region = 19

```

Iteration 0: log likelihood = -698.07521
Iteration 1: log likelihood = -646.47421
Iteration 2: log likelihood = -644.91419
Iteration 3: log likelihood = -644.75345
Iteration 4: log likelihood = -644.72209
Iteration 5: log likelihood = -644.71546
Iteration 6: log likelihood = -644.71397
Iteration 7: log likelihood = -644.71361
Iteration 8: log likelihood = -644.71354
Iteration 9: log likelihood = -644.71352

```

Multinomial logistic regression	Number of obs	=	648
	LR chi2(36)	=	106.72
	Prob > chi2	=	0.0000
Log likelihood = -644.71352	Pseudo R2	=	0.0764

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.043959	.0156963	2.86	0.004	1.013644 1.075181
exper	1.024354	.0216133	1.14	0.254	.9828564 1.067603
male	1.034277	.2871845	0.12	0.903	.6001864 1.782327

age		.9720888	.0166943	-1.65	0.099	.939913	1.005366
pyperstudw~e		.8686703	.1043126	-1.17	0.241	.6864993	1.099183
pyperstudb~k		.8984623	.1127753	-0.85	0.394	.7025168	1.149061
pyperstudl~o		.8839583	.102734	-1.06	0.289	.7038901	1.110091
pyperstude~s		.9922649	.0070184	-1.10	0.272	.9786041	1.006116
pyperstudlep		1.005808	.0068896	0.85	0.398	.9923953	1.019403
pyperstudm~e		.980195	.0096676	-2.03	0.043	.9614288	.9993274
Asian		3666528	6.84e+09	0.01	0.994	0	.
Black		1.176086	.8258247	0.23	0.817	.29699	4.657322
Latino		1.248742	.3530997	0.79	0.432	.7174362	2.173512
Other		1.140751	1.446639	0.10	0.917	.0950043	13.69741
AsianMale		.7378147	1490.704	-0.00	1.000	0	.
BlackMale		.9572349	1.015476	-0.04	0.967	.1196817	7.656133
LatinoMale		1.449862	.5688602	0.95	0.344	.6719762	3.128237
OtherMale		.3495443	1391.122	-0.00	1.000	0	.
-----							
3							
tenure		1.051068	.0163046	3.21	0.001	1.019592	1.083515
exper		1.108667	.0282013	4.06	0.000	1.054749	1.165342
male		.8795642	.2700908	-0.42	0.676	.4818215	1.605643
age		.9408847	.0205399	-2.79	0.005	.9014763	.9820158
pyperstudw~e		1.05091	.1575062	0.33	0.740	.7834134	1.409744
pyperstudb~k		1.095925	.1691389	0.59	0.553	.809862	1.483032
pyperstudl~o		1.052433	.1526877	0.35	0.725	.791957	1.39858
pyperstude~s		.9940124	.0079468	-0.75	0.453	.9785584	1.00971
pyperstudlep		1.020828	.0076475	2.75	0.006	1.005948	1.035927
pyperstudm~e		.9942148	.0087102	-0.66	0.508	.9772888	1.011434
Asian		1.077699	3283.053	0.00	1.000	0	.
Black		.2182266	.2517029	-1.32	0.187	.0227578	2.092593
Latino		1.099206	.3391181	0.31	0.759	.6004453	2.012264
Other		2.46e-06	.001868	-0.02	0.986	0	.
AsianMale		3671687	1.15e+10	0.00	0.996	0	.
BlackMale		2.076204	3.469587	0.44	0.662	.0784891	54.92
LatinoMale		1.509862	.6563446	0.95	0.343	.6440444	3.539638
OtherMale		1.06e+12	2.89e+15	0.01	0.992	0	.
-----							

-----> region = 20

```

Iteration 0: log likelihood = -1860.7019
Iteration 1: log likelihood = -1753.0532
Iteration 2: log likelihood = -1748.7971
Iteration 3: log likelihood = -1748.7118
Iteration 4: log likelihood = -1748.6917
Iteration 5: log likelihood = -1748.6878
Iteration 6: log likelihood = -1748.687
Iteration 7: log likelihood = -1748.687
Iteration 8: log likelihood = -1748.6869

```

Multinomial logistic regression	Number of obs	=	1948
	LR chi2(36)	=	224.03
	Prob > chi2	=	0.0000
Log likelihood = -1748.6869	Pseudo R2	=	0.0602

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----							
1		(base outcome)					
-----							
2							
tenure		1.008133	.0094395	0.87	0.387	.9898004	1.026805
exper		1.016811	.0136701	1.24	0.215	.990368	1.04396
male		1.717896	.2640295	3.52	0.000	1.271079	2.32178
age		.9804312	.0113791	-1.70	0.089	.9583803	1.002989
pyperstudw~e		.9510251	.0483771	-0.99	0.324	.8607812	1.05073
pyperstudb~k		.9552967	.0505257	-0.86	0.387	.861228	1.05964

```

pyperstudl~o | .9527813 .0481646 -0.96 0.339 .8629063 1.052017
pyperstude~s | .9991423 .0052612 -0.16 0.871 .9888836 1.009507
pyperstudlep | .985518 .0066898 -2.15 0.032 .972493 .9987174
pyperstudm~e | .9853018 .0068196 -2.14 0.032 .9720258 .998759
    Asian | 5.02e-06 .002916 -0.02 0.983 0 .
    Black | 1.492566 .5172459 1.16 0.248 .7567479 2.943851
    Latino | 1.354513 .2571203 1.60 0.110 .9336928 1.965
    Other | 4.037222 4.077386 1.38 0.167 .557716 29.22484
    AsianMale | 119743.5 4.98e+08 0.00 0.998 0 .
    BlackMale | .8991752 .4484646 -0.21 0.831 .3383036 2.389912
    LatinoMale | 1.214724 .3238968 0.73 0.466 .7202967 2.048538
    OtherMale | .6102468 .8725259 -0.35 0.730 .0370224 10.05881
-----
3
    tenure | 1.037592 .0096848 3.95 0.000 1.018782 1.056748
    exper | 1.072448 .0167896 4.47 0.000 1.040041 1.105866
    male | 1.601214 .2645703 2.85 0.004 1.158256 2.213576
    age | .9637719 .0129838 -2.74 0.006 .9386572 .9895587
pyperstudw~e | 1.094611 .0639658 1.55 0.122 .9761539 1.227444
pyperstudb~k | 1.046291 .0635612 0.74 0.456 .9288441 1.178589
pyperstudl~o | 1.048532 .0611206 0.81 0.416 .9353281 1.175438
pyperstude~s | 1.042197 .0060561 7.11 0.000 1.030395 1.054135
pyperstudlep | 1.00657 .0056073 1.18 0.240 .9956392 1.01762
pyperstudm~e | .989021 .0067677 -1.61 0.107 .975845 1.002375
    Asian | 1.622802 1.955184 0.40 0.688 .1530104 17.21116
    Black | 1.336639 .4304076 0.90 0.368 .7110839 2.512507
    Latino | 1.175129 .2212497 0.86 0.391 .812501 1.699602
    Other | 2.304818 2.887443 0.67 0.505 .1978171 26.85404
    AsianMale | 4627550 9.58e+09 0.01 0.994 0 .
    BlackMale | .6900111 .3557323 -0.72 0.472 .2512018 1.89535
    LatinoMale | 1.15488 .3134532 0.53 0.596 .6784329 1.965924
    OtherMale | .6293666 1.108897 -0.26 0.793 .0199136 19.89103
-----
by school type
. bysort schooltype: mlogit admin tenure exper male age pyperstudwhite pyperstudblack
pyperstudlatino pyperstudecodis pyperstudlep pyperstudmobile Asi
> an Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1996==1, rrr

-----> schooltype =

note: Other omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -41.497976
Iteration 1: log likelihood = -23.698168
Iteration 2: log likelihood = -18.539897
Iteration 3: log likelihood = -14.066727
Iteration 4: log likelihood = -10.06324
Iteration 5: log likelihood = -9.7847101
Iteration 6: log likelihood = -9.6582531
Iteration 7: log likelihood = -9.6434883
Iteration 8: log likelihood = -9.6392062
Iteration 9: log likelihood = -9.6384389
Iteration 10: log likelihood = -9.6382573
Iteration 11: log likelihood = -9.6382152
Iteration 12: log likelihood = -9.6382058
Iteration 13: log likelihood = -9.6382037
Iteration 14: log likelihood = -9.6382034

Multinomial logistic regression
Number of obs = 48
LR chi2(30) = 63.72
Prob > chi2 = 0.0003
Pseudo R2 = 0.7677
Log likelihood = -9.6382034

-----
admin | RRR Std. Err. z P>|z| [95% Conf. Interval]

```

1		(base outcome)					
2							
	tenure	1.981949	11935.92	0.00	1.000	0	.
	exper	4.902063	83885.54	0.00	1.000	0	.
	male	1.61e+18	1.19e+23	0.00	1.000	0	.
	age	.0499956	909.5627	-0.00	1.000	0	.
	pyperstudw~e	1.29e-06	.0084507	-0.00	0.998	0	.
	pyperstudb~k	9.53e-07	.0063753	-0.00	0.998	0	.
	pyperstudl~o	2.76e-06	.0186352	-0.00	0.998	0	.
	pyperstude~s	.3488098	937.2229	-0.00	1.000	0	.
	pyperstudlep	.0517178	379.6189	-0.00	1.000	0	.
	pyperstudm~e	2.718555	2489.204	0.00	0.999	0	.
	Asian	1.35e+26	1.55e+31	0.00	1.000	0	.
	Black	6.16e+45	7.42e+50	0.00	0.999	0	.
	Latino	6.56e+69	8.27e+74	0.00	0.999	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	9.42e-22	1.23e-16	-0.00	1.000	0	.
	LatinoMale	1.53e-22	1.52e-17	-0.00	1.000	0	.
	OtherMale	(omitted)					

3							
	tenure	1.194901	.1560687	1.36	0.173	.9250269	1.543509
	exper	1.516714	.3951871	1.60	0.110	.9101625	2.527483
	male	44.71401	81.51372	2.08	0.037	1.255167	1592.889
	age	.6331505	.1861339	-1.55	0.120	.3558527	1.126532
	pyperstudw~e	1.320319	.6081074	0.60	0.546	.5353474	3.256282
	pyperstudb~k	1.376322	.6472623	0.68	0.497	.5475371	3.459605
	pyperstudl~o	1.144199	.5432291	0.28	0.777	.4512094	2.901515
	pyperstude~s	1.098641	.0958625	1.08	0.281	.925942	1.303551
	pyperstudlep	1.315362	.2703665	1.33	0.182	.879194	1.967912
	pyperstudm~e	1.126619	.0722476	1.86	0.063	.9935534	1.277505
	Asian	50.22272	335.4284	0.59	0.558	.0001037	2.43e+07
	Black	.2142294	12.57183	-0.03	0.979	2.39e-51	1.92e+49
	Latino	5.23e-12	2.88e-07	-0.00	1.000	0	.
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	1.64e-17	5.84e-10	-0.00	1.000	0	.
	LatinoMale	3.96e+08	2.18e+13	0.00	1.000	0	.
	OtherMale	(omitted)					

-----> schooltype = Both Elem/Sec

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -659.94172  
Iteration 1: log likelihood = -601.10066  
Iteration 2: log likelihood = -595.34771  
Iteration 3: log likelihood = -595.25834  
Iteration 4: log likelihood = -595.2399  
Iteration 5: log likelihood = -595.23694  
Iteration 6: log likelihood = -595.23622  
Iteration 7: log likelihood = -595.23607  
Iteration 8: log likelihood = -595.23604  
Iteration 9: log likelihood = -595.23603

Multinomial logistic regression	Number of obs	=	717
	LR chi2(32)	=	129.41
	Prob > chi2	=	0.0000
Log likelihood = -595.23603	Pseudo R2	=	0.0980

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-------	-----	-----------	---	------	----------------------

1		(base outcome)					
2	tenure	1.018463	.02129	0.88	0.381	.9775781	1.061057
	exper	.9571279	.0229511	-1.83	0.068	.9131854	1.003185
	male	.991894	.3099635	-0.03	0.979	.5376104	1.83005
	age	.9743871	.0206838	-1.22	0.222	.9346792	1.015782
	pyperstudw~e	.9355789	.0383839	-1.62	0.105	.8632931	1.013917
	pyperstudb~k	.9397895	.0402535	-1.45	0.147	.864115	1.022091
	pyperstudl~o	.9490959	.0397173	-1.25	0.212	.8743584	1.030222
	pyperstude~s	.9880376	.0066778	-1.78	0.075	.9750357	1.001213
	pyperstudlep	1.001776	.013426	0.13	0.895	.9758045	1.028439
	pyperstudm~e	.9961839	.0049164	-0.77	0.439	.9865944	1.005867
	Asian	2.487352	9933.136	0.00	1.000	0	.
	Black	2.216913	1.107939	1.59	0.111	.8324276	5.904062
	Latino	2.393724	1.62816	1.28	0.199	.6311098	9.079108
	Other	5.77e-06	.006569	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	.8153186	.5900706	-0.28	0.778	.1973728	3.367964
	LatinoMale	1.010802	.8701419	0.01	0.990	.1870315	5.462823
	OtherMale	(omitted)					
3	tenure	1.014556	.0122313	1.20	0.231	.9908644	1.038815
	exper	1.041758	.0182933	2.33	0.020	1.006513	1.078236
	male	2.202964	.457915	3.80	0.000	1.465797	3.31086
	age	.9536548	.0160731	-2.82	0.005	.9226667	.9856836
	pyperstudw~e	1.201216	.0718313	3.07	0.002	1.068366	1.350585
	pyperstudb~k	1.20174	.073222	3.02	0.003	1.066465	1.354173
	pyperstudl~o	1.194998	.0721628	2.95	0.003	1.06161	1.345144
	pyperstude~s	1.009265	.0050639	1.84	0.066	.9993884	1.019239
	pyperstudlep	.9982231	.0114861	-0.15	0.877	.9759627	1.020991
	pyperstudm~e	.9931578	.0039933	-1.71	0.088	.9853618	1.001016
	Asian	7522063	1.19e+10	0.01	0.992	0	.
	Black	.5678672	.3191674	-1.01	0.314	.1887263	1.708682
	Latino	1.432205	.8970674	0.57	0.566	.4196155	4.88831
	Other	5.689079	6.594997	1.50	0.134	.5865387	55.18071
	AsianMale	(omitted)					
	BlackMale	1.867153	1.278522	0.91	0.362	.4878933	7.145539
	LatinoMale	1.560236	1.132498	0.61	0.540	.3761364	6.471949
	OtherMale	(omitted)					

-----> schooltype = Elementary

Iteration 0: log likelihood = -8880.0861  
Iteration 1: log likelihood = -8325.2589  
Iteration 2: log likelihood = -8319.6123  
Iteration 3: log likelihood = -8319.6075  
Iteration 4: log likelihood = -8319.6075

Multinomial logistic regression	Number of obs	=	8515
	LR chi2(36)	=	1120.96
	Prob > chi2	=	0.0000
Log likelihood = -8319.6075	Pseudo R2	=	0.0631

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.002386	.0050554	0.47	0.637	.992526	1.012343
	exper	1.004137	.0063947	0.65	0.517	.9916818	1.016749
	male	1.933534	.1819612	7.01	0.000	1.607855	2.325181

age		.9834078	.0050556	-3.25	0.001	.9735487	.9933668
pyperstudw~e		.9776554	.0074576	-2.96	0.003	.9631474	.9923818
pyperstudb~k		.9823106	.0078931	-2.22	0.026	.9669615	.9979034
pyperstudl~o		.984386	.0075029	-2.06	0.039	.9697898	.9992018
pyperstude~s		.9836631	.0021757	-7.45	0.000	.9794081	.9879366
pyperstudlep		1.009403	.0024743	3.82	0.000	1.004566	1.014265
pyperstudm~e		1.018049	.0044939	4.05	0.000	1.009279	1.026895
Asian		.381936	.2912453	-1.26	0.207	.0856854	1.70245
Black		1.346077	.1566166	2.55	0.011	1.071599	1.69086
Latino		1.095838	.1168256	0.86	0.391	.8892031	1.35049
Other		1.078272	.4305424	0.19	0.850	.4930067	2.358327
AsianMale		6.308151	6.865126	1.69	0.091	.7473846	53.2427
BlackMale		.756787	.1689845	-1.25	0.212	.4885479	1.172304
LatinoMale		.6826862	.1229236	-2.12	0.034	.4796836	.9715997
OtherMale		.4305493	.3883656	-0.93	0.350	.0734884	2.522474
-----							
3							
tenure		1.015044	.0039991	3.79	0.000	1.007236	1.022913
exper		1.053604	.0058242	9.45	0.000	1.042251	1.065082
male		3.458907	.2585873	16.60	0.000	2.987467	4.004742
age		1.001264	.0046469	0.27	0.785	.9921978	1.010414
pyperstudw~e		1.013749	.0080046	1.73	0.084	.9981808	1.029559
pyperstudb~k		1.003253	.0083171	0.39	0.695	.9870833	1.019687
pyperstudl~o		1.005769	.0080331	0.72	0.471	.9901467	1.021637
pyperstude~s		1.00253	.0020097	1.26	0.208	.9985985	1.006477
pyperstudlep		1.000837	.0021185	0.40	0.693	.9966934	1.004998
pyperstudm~e		1.005834	.0040067	1.46	0.144	.9980112	1.013717
Asian		.6938116	.3359935	-0.75	0.450	.2685521	1.792481
Black		.9849067	.1079208	-0.14	0.890	.7945569	1.220858
Latino		1.384665	.1279464	3.52	0.000	1.155292	1.659579
Other		.4925499	.2173317	-1.60	0.109	.2074282	1.169587
AsianMale		1.454248	1.269241	0.43	0.668	.2628608	8.045468
BlackMale		.6051978	.1205661	-2.52	0.012	.4095652	.8942762
LatinoMale		.4684886	.069479	-5.11	0.000	.3503175	.6265218
OtherMale		.8524088	.6123839	-0.22	0.824	.2085121	3.484693
-----							

-----> schooltype = Middle

Iteration 0: log likelihood = -4943.6252  
Iteration 1: log likelihood = -4648.0619  
Iteration 2: log likelihood = -4639.5346  
Iteration 3: log likelihood = -4638.8665  
Iteration 4: log likelihood = -4638.7634  
Iteration 5: log likelihood = -4638.7384  
Iteration 6: log likelihood = -4638.7332  
Iteration 7: log likelihood = -4638.7321  
Iteration 8: log likelihood = -4638.7319  
Iteration 9: log likelihood = -4638.7318  
Iteration 10: log likelihood = -4638.7318

Multinomial logistic regression	Number of obs	=	4758
	LR chi2(36)	=	609.79
	Prob > chi2	=	0.0000
Log likelihood = -4638.7318	Pseudo R2	=	0.0617

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.002735	.0057074	0.48	0.631	.9916112 1.013984
exper		1.007597	.0074647	1.02	0.307	.9930723 1.022334
male		1.711188	.1471077	6.25	0.000	1.445844 2.025228
age		.9616592	.0060502	-6.21	0.000	.9498739 .9735908



pyperstudw~e		.9671112	.010103	-3.20	0.001	.947511	.9871168
pyperstudb~k		.9671132	.0106121	-3.05	0.002	.9465359	.9881377
pyperstudl~o		.9654904	.0102481	-3.31	0.001	.945612	.9857867
pyperstude~s		.9958297	.002793	-1.49	0.136	.9903705	1.001319
pyperstudlep		1.007315	.0041901	1.75	0.080	.9991358	1.015561
pyperstudm~e		1.002253	.0041887	0.54	0.590	.9940773	1.010497
Asian		.5491703	.6117264	-0.54	0.591	.0618792	4.873816
Black		1.145846	.1720172	0.91	0.364	.8537716	1.537838
Latino		2.020231	.3113629	4.56	0.000	1.493522	2.732689
Other		1.72e-07	.0001984	-0.01	0.989	0	.
AsianMale		4.62e-07	.000457	-0.01	0.988	0	.
BlackMale		1.708197	.3450531	2.65	0.008	1.149732	2.537929
LatinoMale		.9273759	.1775926	-0.39	0.694	.6371631	1.349774
OtherMale		4936168	5.70e+09	0.01	0.989	0	.

3							
tenure		1.007944	.0057152	1.40	0.163	.9968044	1.019208
exper		1.071767	.0096239	7.72	0.000	1.05307	1.090797
male		2.625476	.2460119	10.30	0.000	2.184987	3.154766
age		.9558847	.0076566	-5.63	0.000	.9409952	.9710097
pyperstudw~e		1.018764	.0139702	1.36	0.175	.9917479	1.046517
pyperstudb~k		.9998826	.014461	-0.01	0.994	.9719375	1.028631
pyperstudl~o		1.001727	.0140816	0.12	0.902	.9745039	1.02971
pyperstude~s		1.013584	.0033744	4.05	0.000	1.006991	1.020219
pyperstudlep		.9963419	.0049633	-0.74	0.462	.9866614	1.006117
pyperstudm~e		.9848642	.0059353	-2.53	0.011	.9732996	.9965662
Asian		3.39e-07	.000554	-0.01	0.993	0	.
Black		.9682253	.1964107	-0.16	0.874	.650586	1.440947
Latino		2.154496	.4019422	4.11	0.000	1.494672	3.105601
Other		.4196902	.4462618	-0.82	0.414	.0522199	3.373043
AsianMale		4974008	8.13e+09	0.01	0.992	0	.
BlackMale		1.343966	.3477883	1.14	0.253	.8093129	2.231825
LatinoMale		.7684593	.1690492	-1.20	0.231	.4993094	1.182693
OtherMale		3.982302	4.974411	1.11	0.269	.3442496	46.06754

-----> schooltype = Secondary

Iteration 0: log likelihood = -5795.954  
Iteration 1: log likelihood = -5523.8066  
Iteration 2: log likelihood = -5512.3603  
Iteration 3: log likelihood = -5511.8952  
Iteration 4: log likelihood = -5511.8046  
Iteration 5: log likelihood = -5511.7853  
Iteration 6: log likelihood = -5511.7808  
Iteration 7: log likelihood = -5511.7798  
Iteration 8: log likelihood = -5511.7795  
Iteration 9: log likelihood = -5511.7795

Multinomial logistic regression	Number of obs	=	6104
	LR chi2(36)	=	568.35
	Prob > chi2	=	0.0000
Log likelihood = -5511.7795	Pseudo R2	=	0.0490

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	.9983364	.0044998	-0.37	0.712	.9895558 1.007195
exper	1.000075	.0062879	0.01	0.991	.9878262 1.012475
male	1.399139	.1067558	4.40	0.000	1.204795 1.624832
age	.9809833	.0056019	-3.36	0.001	.970065 .9920244
pyperstudw~e	.9822096	.0074013	-2.38	0.017	.9678098 .9968236
pyperstudb~k	.9824808	.007713	-2.25	0.024	.9674794 .9977149

pyperstudl~o		.9810331	.0074392	-2.53	0.012	.9665603	.9957226
pyperstude~s		.9959495	.0022405	-1.80	0.071	.9915679	1.00035
pyperstudlep		1.006849	.0038909	1.77	0.077	.9992514	1.014504
pyperstudm~e		1.000945	.0029405	0.32	0.748	.9951985	1.006725
Asian		1.464719	1.022094	0.55	0.584	.3730595	5.750829
Black		1.815558	.2719127	3.98	0.000	1.353715	2.434966
Latino		1.522409	.2321159	2.76	0.006	1.129151	2.05263
Other		.5277606	.3316516	-1.02	0.309	.1540036	1.808602
AsianMale		1.920645	1.747975	0.72	0.473	.322679	11.43204
BlackMale		1.294322	.2520734	1.32	0.185	.8836255	1.895905
LatinoMale		1.727149	.3052303	3.09	0.002	1.221522	2.442071
OtherMale		1.716109	1.325804	0.70	0.485	.3775176	7.80104
-----							
3							
tenure		.9775973	.0050058	-4.42	0.000	.9678352	.9874578
exper		1.054041	.0086523	6.41	0.000	1.037219	1.071137
male		2.482047	.2491415	9.06	0.000	2.038772	3.0217
age		.9743076	.0076684	-3.31	0.001	.9593932	.9894539
pyperstudw~e		1.05901	.0143145	4.24	0.000	1.031322	1.087441
pyperstudb~k		1.042413	.0146179	2.96	0.003	1.014153	1.071461
pyperstudl~o		1.037715	.014175	2.71	0.007	1.010301	1.065872
pyperstude~s		1.018775	.0028632	6.62	0.000	1.013179	1.024403
pyperstudlep		1.005518	.0051282	1.08	0.281	.9955175	1.01562
pyperstudm~e		1.014427	.0031136	4.67	0.000	1.008343	1.020548
Asian		1.85e-06	.0014811	-0.02	0.987	0	.
Black		1.251738	.3162964	0.89	0.374	.7628289	2.053998
Latino		1.239245	.2899208	0.92	0.359	.7834636	1.960177
Other		1.30e-06	.0007478	-0.02	0.981	0	.
AsianMale		983382.9	7.88e+08	0.02	0.986	0	.
BlackMale		.9140138	.2827663	-0.29	0.771	.4984465	1.67605
LatinoMale		1.135275	.295834	0.49	0.626	.6812272	1.891952
OtherMale		794221.5	4.55e+08	0.02	0.981	0	.
-----							

### 1996-97

```
. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino Oth
> er AsianMale BlackMale LatinoMale OtherMale if validcert1997a==1, rrr
```

```
Iteration 0: log likelihood = -21210.402
Iteration 1: log likelihood = -20217.384
Iteration 2: log likelihood = -20196.217
Iteration 3: log likelihood = -20196.162
Iteration 4: log likelihood = -20196.162
```

```
Multinomial logistic regression      Number of obs   =      20441
                                     LR chi2(36)       =      2028.48
                                     Prob > chi2      =      0.0000
Log likelihood = -20196.162          Pseudo R2       =      0.0478
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.00246	.0027838	0.88	0.376	.9970189 1.007931
exper		.9986258	.0035921	-0.38	0.702	.9916102 1.005691
pyperstudlep		1.009195	.0016405	5.63	0.000	1.005985 1.012415
pyperstude~s		.9945864	.0011657	-4.63	0.000	.9923043 .9968738
pyperstudw~e		.9714019	.0045952	-6.13	0.000	.9624372 .9804501
pyperstudb~k		.9726213	.0048474	-5.57	0.000	.9631669 .9821685
pyperstudl~o		.9729043	.0046522	-5.74	0.000	.9638288 .9820653
pyperstudm~e		.9985073	.0016216	-0.92	0.358	.9953341 1.001691
male		1.382466	.0609109	7.35	0.000	1.268092 1.507155
age		.9776005	.003037	-7.29	0.000	.9716662 .983571

Asian		.719315	.3094237	-0.77	0.444	.3095726	1.671382
Black		1.35135	.101766	4.00	0.000	1.165913	1.566279
Latino		1.313631	.0936884	3.82	0.000	1.142262	1.51071
Other		.470598	.1713537	-2.07	0.038	.2305215	.9607021
AsianMale		1.553764	.9564497	0.72	0.474	.4649556	5.192288
BlackMale		1.488147	.1630958	3.63	0.000	1.200486	1.844738
LatinoMale		1.319657	.125286	2.92	0.003	1.095594	1.589545
OtherMale		1.920204	.9507005	1.32	0.188	.727636	5.067346
-----							
3							
tenure		1.007059	.0025021	2.83	0.005	1.002167	1.011975
exper		1.05063	.0039087	13.28	0.000	1.042997	1.058319
pyperstudlep		1.006987	.0016215	4.32	0.000	1.003814	1.01017
pyperstude~s		1.026249	.0012687	20.96	0.000	1.023766	1.028739
pyperstudw~e		1.009272	.0055925	1.67	0.096	.9983703	1.020293
pyperstudb~k		.9864713	.0057711	-2.33	0.020	.9752248	.9978476
pyperstudl~o		.9838825	.0055497	-2.88	0.004	.9730652	.9948201
pyperstudm~e		.9964177	.0017932	-1.99	0.046	.9929093	.9999385
male		1.636543	.0680308	11.85	0.000	1.508492	1.775463
age		.9847931	.003291	-4.59	0.000	.9783639	.9912646
Asian		.4766782	.2228892	-1.58	0.113	.1906397	1.191893
Black		.9367841	.0779482	-0.78	0.433	.7958157	1.102723
Latino		1.333254	.0961207	3.99	0.000	1.157566	1.535607
Other		.5124044	.1814385	-1.89	0.059	.2559816	1.025692
AsianMale		4.050835	2.40854	2.35	0.019	1.263102	12.99124
BlackMale		1.195798	.1477058	1.45	0.148	.9386785	1.523347
LatinoMale		1.022215	.0986177	0.23	0.820	.8461026	1.234985
OtherMale		2.299282	1.077965	1.78	0.076	.9173315	5.763127

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=20441)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
tenure		8.038	2	0.018
exper		205.757	2	0.000
pyperstudlep		38.005	2	0.000
pyperstude~s		602.008	2	0.000
pyperstudw~e		51.629	2	0.000
pyperstudb~k		31.212	2	0.000
pyperstudl~o		34.093	2	0.000
pyperstudm~e		4.217	2	0.121
male		156.336	2	0.000
age		61.405	2	0.000
Asian		3.014	2	0.222
Black		19.496	2	0.000
Latino		23.304	2	0.000
Other		7.494	2	0.024
AsianMale		6.001	2	0.050
BlackMale		13.243	2	0.001
LatinoMale		9.394	2	0.009
OtherMale		4.159	2	0.125

\*\*\*\* Wald tests for independent variables (N=20441)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
tenure		8.018	2	0.018
exper		194.441	2	0.000
pyperstudlep		37.739	2	0.000
pyperstude~s		559.974	2	0.000

```

pyperstudw~e |      52.826    2    0.000
pyperstudb~k |      31.335    2    0.000
pyperstudl~o |      34.134    2    0.000
pyperstudm~e |       4.146    2    0.126
      male |    155.488    2    0.000
      age |     60.535    2    0.000
      Asian |       2.750    2    0.253
      Black |     19.869    2    0.000
      Latino |     23.324    2    0.000
      Other |       6.840    2    0.033
      AsianMale |     5.545    2    0.063
      BlackMale |    13.201    2    0.001
      LatinoMale |     9.386    2    0.009
      OtherMale |     4.042    2    0.132
-----

```

\*\*\*\* Hausman tests of IIA assumption (N=20441)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

```

Omitted |      chi2    df    P>chi2    evidence
-----+-----
      2 |    -120.540    18      ---      ---
      3 |     -13.512    18      ---      ---
-----

```

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=20441)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

```

Omitted |    lnL(full)    lnL(omit)      chi2    df    P>chi2    evidence
-----+-----
      2 |   -4713.991   -4701.535   24.912    19     0.163    for Ho
      3 |   -4588.324   -4574.977   26.692    19     0.112    for Ho
-----

```

\*\*\*\* Wald tests for combining alternatives (N=20441)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

```

Alternatives tested|      chi2    df    P>chi2
-----+-----
      2-      3 | 1216.471    18    0.000
      2-      1 |  481.259    18    0.000
      3-      1 | 1165.462    18    0.000
-----

```

\*\*\*\* LR tests for combining alternatives (N=20441)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

```

Alternatives tested|      chi2    df    P>chi2
-----+-----
      2-      3 | 1355.258    18    0.000
      2-      1 |  493.511    18    0.000
      3-      1 | 1308.557    18    0.000
-----

```

by region

```

. bysort region: mlogit admin tenure exper pyperstudlep pyperstudcodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1997a==1, rrr

```

-----> region = 1

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1420.1443  
 Iteration 1: log likelihood = -1337.0725  
 Iteration 2: log likelihood = -1334.6563  
 Iteration 3: log likelihood = -1334.4806  
 Iteration 4: log likelihood = -1334.4418  
 Iteration 5: log likelihood = -1334.4338  
 Iteration 6: log likelihood = -1334.4325  
 Iteration 7: log likelihood = -1334.4322  
 Iteration 8: log likelihood = -1334.4322  
 Iteration 9: log likelihood = -1334.4321

Multinomial logistic regression	Number of obs	=	1331
	LR chi2(34)	=	171.42
	Prob > chi2	=	0.0000
Log likelihood = -1334.4321	Pseudo R2	=	0.0604

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.003976	.0105133	0.38	0.705	.9835806	1.024795
exper	1.036955	.016156	2.33	0.020	1.005768	1.069108
pyperstudlep	.9994065	.0039196	-0.15	0.880	.9917538	1.007118
pyperstude~s	.9962885	.0063286	-0.59	0.558	.9839615	1.00877
pyperstudw~e	1.119111	.1352466	0.93	0.352	.8830874	1.418216
pyperstudb~k	1.145232	.1540419	1.01	0.313	.8798342	1.490686
pyperstudl~o	1.114601	.1275195	0.95	0.343	.8907076	1.394775
pyperstudm~e	1.001952	.0072439	0.27	0.787	.9878548	1.016251
male	1.557083	.450914	1.53	0.126	.8827	2.746694
age	.9528228	.0131561	-3.50	0.000	.927383	.9789603
Asian	7.80e-07	.0010608	-0.01	0.992	0	.
Black	9.77e-07	.0018777	-0.01	0.994	0	.
Latino	1.176582	.2732583	0.70	0.484	.7463328	1.854862
Other	5.28e-07	.0007208	-0.01	0.992	0	.
AsianMale	(omitted)					
BlackMale	1688964	3.24e+09	0.01	0.994	0	.
LatinoMale	1.302385	.425009	0.81	0.418	.6870145	2.468955
OtherMale	2612769	3.57e+09	0.01	0.991	0	.
3						
tenure	1.003604	.0101597	0.36	0.722	.9838873	1.023715
exper	1.111236	.0205108	5.71	0.000	1.071754	1.152172
pyperstudlep	1.014907	.0041039	3.66	0.000	1.006895	1.022983
pyperstude~s	1.019546	.0076336	2.59	0.010	1.004693	1.034618
pyperstudw~e	1.142667	.1377865	1.11	0.269	.902151	1.447306
pyperstudb~k	1.146459	.1565768	1.00	0.317	.8772153	1.498341
pyperstudl~o	1.081551	.1232985	0.69	0.492	.8649854	1.352338
pyperstudm~e	1.012017	.0076736	1.58	0.115	.9970879	1.027169
male	.6368285	.2183983	-1.32	0.188	.3251664	1.247209
age	.9462251	.0158787	-3.29	0.001	.9156096	.9778643
Asian	2.37e-07	.000329	-0.01	0.991	0	.
Black	2.68e-06	.0053728	-0.01	0.995	0	.
Latino	.9523497	.2192902	-0.21	0.832	.6064525	1.495533
Other	1.29e-07	.0001625	-0.01	0.990	0	.
AsianMale	(omitted)					
BlackMale	.7518583	1724.81	-0.00	1.000	0	.
LatinoMale	2.7961	1.057391	2.72	0.007	1.332468	5.867439
OtherMale	1.82e+07	2.29e+10	0.01	0.989	0	.

-----> region = 2

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -811.02632  
 Iteration 1: log likelihood = -757.02525  
 Iteration 2: log likelihood = -755.23998  
 Iteration 3: log likelihood = -754.11612  
 Iteration 4: log likelihood = -754.10577  
 Iteration 5: log likelihood = -754.10329  
 Iteration 6: log likelihood = -754.1028  
 Iteration 7: log likelihood = -754.10271  
 Iteration 8: log likelihood = -754.10269

Multinomial logistic regression	Number of obs	=	845
	LR chi2(32)	=	113.85
	Prob > chi2	=	0.0000
Log likelihood = -754.10269	Pseudo R2	=	0.0702

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005541	.016484	0.34	0.736	.9737465	1.038374
	exper	.9793046	.0207122	-0.99	0.323	.9395393	1.020753
pyperstudlep		.9922119	.0216153	-0.36	0.720	.9507384	1.035495
pyperstude~s		1.001104	.0062976	0.18	0.861	.9888368	1.013524
pyperstudw~e		.9618467	.0663024	-0.56	0.573	.8402926	1.100985
pyperstudb~k		.9768717	.0702637	-0.33	0.745	.848424	1.124766
pyperstudl~o		.9632869	.0635364	-0.57	0.571	.8464712	1.096224
pyperstudm~e		.9957227	.0084131	-0.51	0.612	.979369	1.012349
	male	1.357768	.3638151	1.14	0.254	.8030554	2.29565
	age	.9816961	.0164296	-1.10	0.270	.950017	1.014431
	Asian	(omitted)					
	Black	1.680076	1.105703	0.79	0.430	.4625203	6.102773
	Latino	.8344169	.2328396	-0.65	0.517	.4829029	1.441805
	Other	7.908229	9.381571	1.74	0.081	.773205	80.88422
	AsianMale	(omitted)					
	BlackMale	1.339136	1.272261	0.31	0.759	.2080322	8.620223
	LatinoMale	1.701649	.6848838	1.32	0.187	.7731726	3.745102
	OtherMale	.6508821	1.127583	-0.25	0.804	.0218216	19.4141
3							
	tenure	.9856097	.0124496	-1.15	0.251	.9615086	1.010315
	exper	1.099789	.0226079	4.63	0.000	1.056359	1.145004
pyperstudlep		1.00526	.0165152	0.32	0.749	.9734065	1.038156
pyperstude~s		1.028846	.0063071	4.64	0.000	1.016559	1.041283
pyperstudw~e		1.062808	.0691057	0.94	0.349	.935639	1.207262
pyperstudb~k		1.03959	.0720736	0.56	0.575	.9075055	1.190899
pyperstudl~o		1.038565	.0647413	0.61	0.544	.9191194	1.173532
pyperstudm~e		.9943553	.0080725	-0.70	0.486	.9786588	1.010304
	male	1.244437	.3239225	0.84	0.401	.7471505	2.072706
	age	.9697051	.0177842	-1.68	0.093	.9354677	1.005196
	Asian	(omitted)					
	Black	.966072	.7184557	-0.05	0.963	.2248959	4.149899
	Latino	1.051773	.2663097	0.20	0.842	.6403198	1.727614
	Other	7.05e-06	.0062514	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	2.246057	2.304394	0.79	0.430	.3006843	16.77763
	LatinoMale	1.8383	.671036	1.67	0.095	.8988797	3.759509
	OtherMale	2.164626	2423.712	0.00	0.999	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -335.00463  
Iteration 1: log likelihood = -303.05839  
Iteration 2: log likelihood = -302.60498  
Iteration 3: log likelihood = -301.47464  
Iteration 4: log likelihood = -301.38675  
Iteration 5: log likelihood = -301.38209  
Iteration 6: log likelihood = -301.38124  
Iteration 7: log likelihood = -301.38105  
Iteration 8: log likelihood = -301.381  
Iteration 9: log likelihood = -301.38099

Multinomial logistic regression	Number of obs	=	329
	LR chi2(30)	=	67.25
	Prob > chi2	=	0.0001
Log likelihood = -301.38099	Pseudo R2	=	0.1004

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.007944	.0257796	0.31	0.757	.958662	1.059759
exper	.9500072	.0349703	-1.39	0.164	.8838807	1.021081
pyperstudlep	.9853841	.0548582	-0.26	0.791	.8835224	1.09899
pyperstude~s	.9888146	.0151738	-0.73	0.464	.9595172	1.019006
pyperstudw~e	.9931697	.0586211	-0.12	0.908	.8846713	1.114975
pyperstudb~k	1.040226	.0588679	0.70	0.486	.931016	1.162248
pyperstudl~o	1.012834	.0591658	0.22	0.827	.9032638	1.135697
pyperstudm~e	1.006552	.018479	0.36	0.722	.9709779	1.04343
male	.8242365	.3246442	-0.49	0.624	.3808789	1.783679
age	1.003277	.031667	0.10	0.917	.9430912	1.067303
Asian	(omitted)					
Black	2.914354	2.534996	1.23	0.219	.5298349	16.03038
Latino	.6196537	.5463616	-0.54	0.587	.1100593	3.488761
Other	3.34e-06	.0025098	-0.02	0.987	0	.
AsianMale	(omitted)					
BlackMale	1.260653	2.049236	0.14	0.887	.0521119	30.49677
LatinoMale	3.615938	4.035047	1.15	0.249	.4058447	32.21677
OtherMale	(omitted)					
3						
tenure	.9830978	.0175268	-0.96	0.339	.9493392	1.018057
exper	1.024768	.0302015	0.83	0.406	.9672511	1.085705
pyperstudlep	1.075437	.0395229	1.98	0.048	1.000698	1.155759
pyperstude~s	1.034549	.012234	2.87	0.004	1.010846	1.058807
pyperstudw~e	1.086279	.0450865	1.99	0.046	1.00141	1.178341
pyperstudb~k	1.068056	.0434892	1.62	0.106	.986131	1.156787
pyperstudl~o	1.051101	.0437477	1.20	0.231	.9687613	1.140439
pyperstudm~e	1.013981	.0162031	0.87	0.385	.9827153	1.046241
male	3.214393	.9948877	3.77	0.000	1.752441	5.895961
age	1.005262	.0278854	0.19	0.850	.9520664	1.061429
Asian	(omitted)					
Black	6.065295	4.799316	2.28	0.023	1.286227	28.60134
Latino	.2948143	.3264385	-1.10	0.270	.0336543	2.582599
Other	1.255943	1.843037	0.16	0.877	.0707737	22.28783
AsianMale	(omitted)					
BlackMale	4.87e-08	.000095	-0.01	0.993	0	.
LatinoMale	4.89525	6.135177	1.27	0.205	.419731	57.09246
OtherMale	(omitted)					

-----> region = 4

Iteration 0: log likelihood = -4266.707  
 Iteration 1: log likelihood = -4006.6008  
 Iteration 2: log likelihood = -3995.9817  
 Iteration 3: log likelihood = -3995.5998  
 Iteration 4: log likelihood = -3995.5319  
 Iteration 5: log likelihood = -3995.517  
 Iteration 6: log likelihood = -3995.5135  
 Iteration 7: log likelihood = -3995.5127  
 Iteration 8: log likelihood = -3995.5126  
 Iteration 9: log likelihood = -3995.5125  
 Iteration 10: log likelihood = -3995.5125

Multinomial logistic regression	Number of obs	=	4159
	LR chi2(36)	=	542.39
	Prob > chi2	=	0.0000
Log likelihood = -3995.5125	Pseudo R2	=	0.0636

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.004385	.0062146	0.71	0.479	.9922783	1.01664
	exper	1.033146	.0091941	3.66	0.000	1.015283	1.051325
pyperstudlep		1.006427	.004681	1.38	0.168	.9972944	1.015644
pyperstude~s		.9971307	.002432	-1.18	0.239	.9923753	1.001909
pyperstudw~e		.9896475	.0070933	-1.45	0.147	.9758421	1.003648
pyperstudb~k		.9836686	.0075875	-2.13	0.033	.9689092	.9986528
pyperstudl~o		.9904551	.0072234	-1.32	0.188	.9763981	1.004714
pyperstudm~e		.995017	.0034089	-1.46	0.145	.988358	1.001721
	male	1.311477	.1261842	2.82	0.005	1.086081	1.583651
	age	.950029	.007205	-6.76	0.000	.936012	.964256
	Asian	.9818388	.7049605	-0.03	0.980	.2403683	4.010543
	Black	.9216643	.1163771	-0.65	0.518	.7196028	1.180464
	Latino	1.140118	.2145348	0.70	0.486	.7884613	1.648616
	Other	.4059171	.319174	-1.15	0.252	.086923	1.895571
	AsianMale	2.27e-07	.0002377	-0.01	0.988	0	.
	BlackMale	1.802346	.3311927	3.21	0.001	1.257258	2.583758
	LatinoMale	1.646061	.4899195	1.67	0.094	.9185478	2.949785
	OtherMale	8.06e-07	.001014	-0.01	0.991	0	.
3							
	tenure	1.033379	.0069087	4.91	0.000	1.019926	1.047008
	exper	1.07917	.0114805	7.16	0.000	1.056901	1.101907
pyperstudlep		1.014023	.005416	2.61	0.009	1.003463	1.024694
pyperstude~s		1.022266	.0032514	6.92	0.000	1.015913	1.028658
pyperstudw~e		1.007845	.0090104	0.87	0.382	.9903385	1.02566
pyperstudb~k		.9895781	.0098271	-1.05	0.291	.9705036	1.009027
pyperstudl~o		.9795357	.0091439	-2.21	0.027	.9617769	.9976224
pyperstudm~e		.9921181	.0043142	-1.82	0.069	.9836983	1.00061
	male	1.311692	.1417237	2.51	0.012	1.061359	1.621068
	age	.9650314	.0088644	-3.88	0.000	.9478129	.9825626
	Asian	.3298363	.3664826	-1.00	0.318	.0373701	2.911201
	Black	.5201144	.0821976	-4.14	0.000	.3815725	.7089583
	Latino	2.071667	.3944167	3.83	0.000	1.426472	3.008687
	Other	.7489799	.5974203	-0.36	0.717	.1568604	3.576244
	AsianMale	2.388301	3.315599	0.63	0.531	.157179	36.28973
	BlackMale	1.551935	.3575678	1.91	0.056	.9879968	2.437763
	LatinoMale	1.201091	.3873513	0.57	0.570	.6383576	2.259894
	OtherMale	6.269779	7.339525	1.57	0.117	.6321488	62.18494

-----> region = 5



note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -557.76471  
Iteration 1: log likelihood = -502.41296  
Iteration 2: log likelihood = -500.43671  
Iteration 3: log likelihood = -500.27708  
Iteration 4: log likelihood = -500.24937  
Iteration 5: log likelihood = -500.24318  
Iteration 6: log likelihood = -500.24169  
Iteration 7: log likelihood = -500.24138  
Iteration 8: log likelihood = -500.24131  
Iteration 9: log likelihood = -500.2413

Multinomial logistic regression	Number of obs	=	562
	LR chi2(30)	=	115.05
	Prob > chi2	=	0.0000
Log likelihood = -500.2413	Pseudo R2	=	0.1031

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.034183	.0185837	1.87	0.061	.9983937	1.071256
	exper	.9902235	.0279885	-0.35	0.728	.9368588	1.046628
pyperstudlep		.9586842	.0454867	-0.89	0.374	.8735516	1.052113
pyperstude~s		1.008	.0093653	0.86	0.391	.98981	1.026523
pyperstudw~e		.950045	.0379101	-1.28	0.199	.8785739	1.02733
pyperstudb~k		.9444233	.039039	-1.38	0.167	.8709258	1.024123
pyperstudl~o		.9709453	.0477701	-0.60	0.549	.8816902	1.069236
pyperstudm~e		.9552851	.0271116	-1.61	0.107	.9035981	1.009929
	male	2.971787	.8417638	3.85	0.000	1.705742	5.177523
	age	.9601322	.0234247	-1.67	0.095	.9153011	1.007159
	Asian	(omitted)					
	Black	1.901615	.7720166	1.58	0.113	.8581282	4.213984
	Latino	3.999051	5.064563	1.09	0.274	.3341644	47.8579
	Other	6.32e-07	.0010202	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	.9240626	.5036561	-0.14	0.885	.3175104	2.689335
	LatinoMale	.1664144	.2972678	-1.00	0.315	.0050197	5.516977
	OtherMale	(omitted)					
3							
	tenure	1.029916	.0160168	1.90	0.058	.9989974	1.061792
	exper	1.0378	.0277115	1.39	0.165	.9848828	1.093559
pyperstudlep		.9721948	.0408002	-0.67	0.502	.8954283	1.055543
pyperstude~s		1.039847	.0090356	4.50	0.000	1.022287	1.057708
pyperstudw~e		1.027131	.0432333	0.64	0.525	.9457962	1.11546
pyperstudb~k		.9907105	.0431827	-0.21	0.830	.9095885	1.079067
pyperstudl~o		1.03051	.0516267	0.60	0.549	.9341329	1.136831
pyperstudm~e		1.02323	.0173859	1.35	0.177	.989715	1.057879
	male	3.614709	.9781576	4.75	0.000	2.126836	6.143453
	age	.9726164	.0236573	-1.14	0.254	.9273367	1.020107
	Asian	(omitted)					
	Black	1.880066	.8020246	1.48	0.139	.8148088	4.338009
	Latino	8.93e-06	.0067937	-0.02	0.988	0	.
	Other	8.65e-07	.0012949	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	.8765975	.4941951	-0.23	0.815	.2903504	2.646537
	LatinoMale	.13021	140.8307	-0.00	0.998	0	.
	OtherMale	(omitted)					

-----> region = 6

Iteration 0: log likelihood = -705.09939  
 Iteration 1: log likelihood = -652.43356  
 Iteration 2: log likelihood = -650.85219  
 Iteration 3: log likelihood = -650.57824  
 Iteration 4: log likelihood = -650.51778  
 Iteration 5: log likelihood = -650.50341  
 Iteration 6: log likelihood = -650.50032  
 Iteration 7: log likelihood = -650.49968  
 Iteration 8: log likelihood = -650.49954  
 Iteration 9: log likelihood = -650.4995  
 Iteration 10: log likelihood = -650.49949

Multinomial logistic regression	Number of obs	=	679
	LR chi2(36)	=	109.20
	Prob > chi2	=	0.0000
Log likelihood = -650.49949	Pseudo R2	=	0.0774

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9756316	.0155396	-1.55	0.121	.945645	1.006569
	exper	1.013378	.0213001	0.63	0.527	.9724789	1.055997
pyperstudlep		1.047861	.0249568	1.96	0.050	1.00007	1.097935
pyperstude~s		.9991953	.0082245	-0.10	0.922	.9832048	1.015446
pyperstudw~e		1.011065	.0637627	0.17	0.861	.8935074	1.144089
pyperstudb~k		1.006248	.0652354	0.10	0.923	.8861788	1.142586
pyperstudl~o		1.01889	.0640708	0.30	0.766	.9007441	1.152533
pyperstudm~e		.9864187	.0140137	-0.96	0.336	.9593313	1.014271
	male	2.035134	.4593622	3.15	0.002	1.307564	3.167545
	age	.9725966	.0182093	-1.48	0.138	.9375538	1.008949
	Asian	2.61e-07	.0009688	-0.00	0.997	0	.
	Black	1.094853	.6821538	0.15	0.884	.3228571	3.712798
	Latino	.308789	.3580958	-1.01	0.311	.0318084	2.997654
	Other	2.92e-07	.0010828	-0.00	0.997	0	.
	AsianMale	1.60e+13	6.56e+16	0.01	0.994	0	.
	BlackMale	1.891867	1.543911	0.78	0.435	.3821584	9.365644
	LatinoMale	2.906298	4.307952	0.72	0.472	.1590815	53.09586
	OtherMale	1.296394	5618.281	0.00	1.000	0	.
3							
	tenure	.9979139	.0137127	-0.15	0.879	.9713963	1.025156
	exper	1.075422	.0236159	3.31	0.001	1.030118	1.122719
pyperstudlep		.9999667	.0240239	-0.00	0.999	.953972	1.048179
pyperstude~s		1.036713	.0074856	4.99	0.000	1.022145	1.051489
pyperstudw~e		.9247373	.0481695	-1.50	0.133	.8349864	1.024135
pyperstudb~k		.9112786	.0484655	-1.75	0.081	.8210712	1.011397
pyperstudl~o		.9149463	.0474161	-1.72	0.086	.8265764	1.012764
pyperstudm~e		1.009348	.0114809	0.82	0.413	.9870949	1.032103
	male	2.601721	.5493069	4.53	0.000	1.720064	3.93529
	age	.957275	.0190447	-2.19	0.028	.9206664	.9953393
	Asian	3.92e-07	.0012853	-0.00	0.996	0	.
	Black	1.102708	.6588665	0.16	0.870	.341876	3.556742
	Latino	4.11e-07	.0004803	-0.01	0.990	0	.
	Other	3.05e-07	.0010002	-0.00	0.996	0	.
	AsianMale	3.78e+12	1.41e+16	0.01	0.994	0	.
	BlackMale	.4008795	.3531773	-1.04	0.299	.0713018	2.253863
	LatinoMale	2450589	2.86e+09	0.01	0.990	0	.
	OtherMale	3306585	1.08e+10	0.00	0.996	0	.

-----> region = 7

note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1006.8005  
Iteration 1: log likelihood = -914.91684  
Iteration 2: log likelihood = -909.44877  
Iteration 3: log likelihood = -909.32746  
Iteration 4: log likelihood = -909.30399  
Iteration 5: log likelihood = -909.30021  
Iteration 6: log likelihood = -909.2998  
Iteration 7: log likelihood = -909.29971  
Iteration 8: log likelihood = -909.29969

Multinomial logistic regression	Number of obs	=	1055
	LR chi2(34)	=	195.00
	Prob > chi2	=	0.0000
Log likelihood = -909.29969	Pseudo R2	=	0.0968

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.009498	.0151162	0.63	0.528	.9803014	1.039565
	exper	.9851564	.0181004	-0.81	0.416	.9503115	1.021279
pypersistudlep		.9995332	.0274163	-0.02	0.986	.9472172	1.054739
pypersistude~s		.9905121	.0076629	-1.23	0.218	.9756064	1.005646
pypersistudw~e		.8853319	.1137046	-0.95	0.343	.6883113	1.138747
pypersistudb~k		.8887553	.1145568	-0.91	0.360	.6903446	1.144191
pypersistudl~o		.9026237	.1177598	-0.79	0.432	.6989655	1.165622
pypersistudm~e		.9678459	.0156056	-2.03	0.043	.9377377	.9989208
	male	2.604791	.6074775	4.11	0.000	1.649147	4.114208
	age	.9847952	.0160985	-0.94	0.349	.9537427	1.016859
	Asian	1.75e+08	6.95e+11	0.00	0.996	0	.
	Black	2.086132	.7879802	1.95	0.052	.9950003	4.373813
	Latino	.8896882	1.011378	-0.10	0.918	.0958541	8.257812
	Other	5.33e-06	.0044222	-0.01	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	1.967093	1.032976	1.29	0.198	.7028054	5.505729
	LatinoMale	3.787934	5.800632	0.87	0.384	.1883299	76.18778
	OtherMale	928239.8	7.71e+08	0.02	0.987	0	.
3							
	tenure	.9871173	.0104469	-1.23	0.221	.9668527	1.007807
	exper	1.078081	.0165481	4.90	0.000	1.04613	1.111007
pypersistudlep		1.011084	.0226482	0.49	0.623	.9676542	1.056462
pypersistude~s		1.047125	.0063091	7.64	0.000	1.034832	1.059564
pypersistudw~e		.8987077	.1003131	-0.96	0.339	.7221176	1.118482
pypersistudb~k		.87273	.0978465	-1.21	0.225	.7005629	1.087208
pypersistudl~o		.8674097	.0987886	-1.25	0.212	.6938756	1.084344
pypersistudm~e		.9959592	.0097591	-0.41	0.679	.9770142	1.015272
	male	1.878691	.3172436	3.73	0.000	1.349329	2.615729
	age	.9582194	.0138494	-2.95	0.003	.931456	.9857517
	Asian	2.585764	17586.25	0.00	1.000	0	.
	Black	.6289369	.2196888	-1.33	0.184	.3171617	1.247192
	Latino	.2424903	.2699619	-1.27	0.203	.0273565	2.149454
	Other	5.47e-06	.0030861	-0.02	0.983	0	.
	AsianMale	(omitted)					
	BlackMale	3.896561	1.944898	2.72	0.006	1.46494	10.36438
	LatinoMale	2.523013	4.238187	0.55	0.582	.0937694	67.88564
	OtherMale	449819.7	2.54e+08	0.02	0.982	0	.

-----> region = 8

note: AsianMale omitted because of collinearity

note: LatinoMale omitted because of collinearity

Iteration 0: log likelihood = -304.76324  
 Iteration 1: log likelihood = -277.65791  
 Iteration 2: log likelihood = -276.24136  
 Iteration 3: log likelihood = -276.0533  
 Iteration 4: log likelihood = -276.0237  
 Iteration 5: log likelihood = -276.02055  
 Iteration 6: log likelihood = -276.0199  
 Iteration 7: log likelihood = -276.01977  
 Iteration 8: log likelihood = -276.01974  
 Iteration 9: log likelihood = -276.01973

Multinomial logistic regression

Number of obs = 316  
 LR chi2(32) = 57.49  
 Prob > chi2 = 0.0037  
 Pseudo R2 = 0.0943

Log likelihood = -276.01973

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.028235	.0283293	1.01	0.312	.9741836	1.085286
	exper	1.00839	.0471108	0.18	0.858	.9201558	1.105085
pyperstudlep		.9683541	.0726301	-0.43	0.668	.8359705	1.121702
pyperstude~s		1.001769	.0161792	0.11	0.913	.9705555	1.033987
pyperstudw~e		.8633224	.2064919	-0.61	0.539	.5402319	1.37964
pyperstudb~k		.8719675	.2095276	-0.57	0.569	.5444558	1.39649
pyperstudl~o		.8694181	.2154577	-0.56	0.572	.5349157	1.413097
pyperstudm~e		.9857303	.0289365	-0.49	0.624	.9306166	1.044108
	male	2.673866	1.182022	2.22	0.026	1.124225	6.359541
	age	.9629171	.0389478	-0.93	0.350	.8895283	1.042361
	Asian	7.43e+08	1.11e+13	0.00	0.999	0	.
	Black	6.077569	4.842559	2.26	0.024	1.274968	28.9708
	Latino	9.52e-07	.0022308	-0.01	0.995	0	.
	Other	5.53e-07	.0011167	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.4580874	.49522	-0.72	0.470	.0550489	3.81196
	LatinoMale	(omitted)					
	OtherMale	.8862597	2554.831	-0.00	1.000	0	.
3							
	tenure	.9788113	.0176874	-1.19	0.236	.9447514	1.014099
	exper	1.090656	.0342801	2.76	0.006	1.025496	1.159956
pyperstudlep		1.03849	.0539464	0.73	0.467	.9379615	1.149793
pyperstude~s		1.03265	.0109354	3.03	0.002	1.011437	1.054307
pyperstudw~e		.9602943	.1808688	-0.22	0.830	.6638698	1.389075
pyperstudb~k		.9400511	.1776113	-0.33	0.744	.6491233	1.361369
pyperstudl~o		.9155905	.178442	-0.45	0.651	.6248967	1.341511
pyperstudm~e		1.024839	.013961	1.80	0.072	.9978383	1.052571
	male	2.12499	.6107185	2.62	0.009	1.209826	3.732422
	age	.9602578	.0264719	-1.47	0.141	.9097507	1.013569
	Asian	.9427015	22477.25	-0.00	1.000	0	.
	Black	1.760721	1.333892	0.75	0.455	.3988732	7.772239
	Latino	3.65e-07	.000554	-0.01	0.992	0	.
	Other	7.92e-07	.0009773	-0.01	0.991	0	.
	AsianMale	(omitted)					
	BlackMale	.9508013	.9641577	-0.05	0.960	.130296	6.938228
	LatinoMale	(omitted)					
	OtherMale	1603080	1.98e+09	0.01	0.991	0	.

-----> region = 9

note: Asian omitted because of collinearity

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -201.36799  
Iteration 1: log likelihood = -175.09499  
Iteration 2: log likelihood = -172.50087  
Iteration 3: log likelihood = -172.22861  
Iteration 4: log likelihood = -172.18774  
Iteration 5: log likelihood = -172.17828  
Iteration 6: log likelihood = -172.17599  
Iteration 7: log likelihood = -172.17554  
Iteration 8: log likelihood = -172.17546  
Iteration 9: log likelihood = -172.17545  
Iteration 10: log likelihood = -172.17544

Multinomial logistic regression                      Number of obs    =        209  
   LR chi2(30)       =        58.39  
   Prob > chi2        =        0.0014  
Log likelihood = -172.17544                           Pseudo R2        =        0.1450

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.022508	.0448425	0.51	0.612	.9382895	1.114285
	exper	.9584199	.0595943	-0.68	0.495	.848454	1.082638
pyperstudlep		.8870846	.1991861	-0.53	0.594	.571263	1.377507
pyperstude~s		.9679303	.0240385	-1.31	0.189	.9219439	1.01621
pyperstudw~e		1.110388	.2169995	0.54	0.592	.7570544	1.628629
pyperstudb~k		1.200609	.258328	0.85	0.395	.787505	1.830415
pyperstudl~o		1.109106	.2258198	0.51	0.611	.7441562	1.653036
pyperstudm~e		.9756897	.0414935	-0.58	0.563	.897661	1.060501
	male	3.127398	1.923206	1.85	0.064	.9369864	10.43838
	age	.9467871	.0429588	-1.21	0.228	.8662246	1.034842
	Asian	(omitted)					
	Black	39.29414	58.72593	2.46	0.014	2.099853	735.3036
	Latino	343.3032	2119765	0.00	0.999	0	.
	Other	1.43e+09	5.38e+13	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	4.45e-09	.0000136	-0.01	0.995	0	.
	LatinoMale	5.49e-10	3.73e-06	-0.00	0.997	0	.
	OtherMale	(omitted)					
3							
	tenure	.9879643	.0221868	-0.54	0.590	.9454221	1.032421
	exper	1.08775	.0400942	2.28	0.022	1.011938	1.169241
pyperstudlep		1.002908	.0775484	0.04	0.970	.8618727	1.167022
pyperstude~s		1.026952	.0141913	1.92	0.054	.9995109	1.055147
pyperstudw~e		1.158822	.142852	1.20	0.232	.9100938	1.475528
pyperstudb~k		1.169332	.1585465	1.15	0.249	.8964493	1.525281
pyperstudl~o		1.14633	.1421234	1.10	0.271	.8990354	1.461648
pyperstudm~e		1.023196	.0246674	0.95	0.342	.9759734	1.072704
	male	2.864527	.9872494	3.05	0.002	1.457771	5.62881
	age	.9606868	.0297736	-1.29	0.196	.9040685	1.020851
	Asian	(omitted)					
	Black	2.183511	2.924682	0.58	0.560	.1581327	30.15011
	Latino	4493341	1.24e+10	0.01	0.996	0	.
	Other	2.242403	119054.6	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	6.88e-08	.0001089	-0.01	0.992	0	.
	LatinoMale	1.52e-07	.0004188	-0.01	0.995	0	.
	OtherMale	(omitted)					

-----> region = 10

```

Iteration 0: log likelihood = -2813.0248
Iteration 1: log likelihood = -2656.9782
Iteration 2: log likelihood = -2654.1581
Iteration 3: log likelihood = -2654.1424
Iteration 4: log likelihood = -2654.1423

```

```

Multinomial logistic regression
Number of obs   =      2693
LR chi2(36)     =      317.76
Prob > chi2     =      0.0000
Pseudo R2      =      0.0565

Log likelihood = -2654.1423

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005076	.0075806	0.67	0.502	.9903275	1.020044
	exper	1.025306	.0109746	2.33	0.020	1.00402	1.047043
pyperstudlep		.998781	.0069273	-0.18	0.860	.9852956	1.012451
pyperstude~s		.9947497	.0035268	-1.48	0.138	.9878612	1.001686
pyperstudw~e		.9926796	.0121375	-0.60	0.548	.9691733	1.016756
pyperstudb~k		.9955678	.0122602	-0.36	0.718	.9718258	1.01989
pyperstudl~o		.9989584	.0123336	-0.08	0.933	.975075	1.023427
pyperstudm~e		.994576	.0055658	-0.97	0.331	.9837268	1.005545
	male	1.668313	.19918	4.29	0.000	1.320239	2.108155
	age	.9647821	.008821	-3.92	0.000	.9476473	.9822267
	Asian	.4462935	.4811491	-0.75	0.454	.0539436	3.692335
	Black	2.134556	.3794066	4.27	0.000	1.506645	3.024157
	Latino	2.814039	.7053549	4.13	0.000	1.721756	4.599267
	Other	.2153163	.2238536	-1.48	0.140	.0280624	1.652072
	AsianMale	2.337175	2.913682	0.68	0.496	.2030137	26.9065
	BlackMale	1.027469	.2676782	0.10	0.917	.6166124	1.712086
	LatinoMale	.763369	.2824746	-0.73	0.466	.3696227	1.57656
	OtherMale	4.091924	5.125501	1.12	0.261	.3513352	47.65776
3							
	tenure	1.012482	.0071475	1.76	0.079	.9985695	1.026588
	exper	1.080697	.012409	6.76	0.000	1.056647	1.105294
pyperstudlep		1.009032	.0075002	1.21	0.226	.9944382	1.02384
pyperstude~s		1.019831	.0037371	5.36	0.000	1.012533	1.027182
pyperstudw~e		1.023741	.0129134	1.86	0.063	.9987416	1.049366
pyperstudb~k		1.004758	.0130701	0.36	0.715	.979465	1.030705
pyperstudl~o		.9933468	.0129641	-0.51	0.609	.9682598	1.019084
pyperstudm~e		.9896342	.0064185	-1.61	0.108	.9771338	1.002295
	male	1.505059	.1763349	3.49	0.000	1.196259	1.89357
	age	.9653909	.0096578	-3.52	0.000	.9466463	.9845066
	Asian	.5993798	.4352468	-0.70	0.481	.1444082	2.487782
	Black	1.241586	.2461524	1.09	0.275	.8418226	1.831188
	Latino	2.368346	.6289874	3.25	0.001	1.407286	3.985731
	Other	.1712219	.180406	-1.67	0.094	.021712	1.350262
	AsianMale	4.236418	3.726666	1.64	0.101	.7554733	23.75628
	BlackMale	1.303114	.3833986	0.90	0.368	.7320557	2.31964
	LatinoMale	1.069817	.4250123	0.17	0.865	.491075	2.330619
	OtherMale	8.454405	10.51308	1.72	0.086	.7389429	96.72868

```

-----> region = 11

```

```

note: AsianMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -1977.4489
Iteration 1: log likelihood = -1837.2682
Iteration 2: log likelihood = -1835.4796
Iteration 3: log likelihood = -1835.401
Iteration 4: log likelihood = -1835.3929

```

```

Iteration 5: log likelihood = -1835.3916
Iteration 6: log likelihood = -1835.3914
Iteration 7: log likelihood = -1835.3914
Iteration 8: log likelihood = -1835.3914

```

```

Multinomial logistic regression
Log likelihood = -1835.3914
Number of obs   =      1845
LR chi2(34)     =      284.12
Prob > chi2     =      0.0000
Pseudo R2      =      0.0718

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9687325	.009469	-3.25	0.001	.9503503	.9874703
exper	.9529673	.0094896	-4.84	0.000	.9345484	.9717491
pyperstudlep	1.009129	.0125623	0.73	0.465	.9848054	1.034054
pyperstude~s	1.001839	.0042753	0.43	0.667	.9934943	1.010254
pyperstudw~e	.9520044	.0173308	-2.70	0.007	.9186355	.9865854
pyperstudb~k	.9436113	.0178325	-3.07	0.002	.9092996	.9792177
pyperstudl~o	.936842	.0171689	-3.56	0.000	.9037888	.9711104
pyperstudm~e	.9931315	.0051455	-1.33	0.183	.9830975	1.003268
male	1.032735	.1374963	0.24	0.809	.7955393	1.340652
age	1.02554	.0087976	2.94	0.003	1.008441	1.042929
Asian	.4714733	.5416448	-0.65	0.513	.0496098	4.48071
Black	1.593844	.4554876	1.63	0.103	.9103118	2.790626
Latino	2.148287	.7287548	2.25	0.024	1.10495	4.176784
Other	.8317962	.7320002	-0.21	0.834	.1482312	4.667607
AsianMale	(omitted)					
BlackMale	1.807567	.7431295	1.44	0.150	.8075091	4.046146
LatinoMale	.5951615	.2991772	-1.03	0.302	.2222035	1.594112
OtherMale	2.21e-06	.0014913	-0.02	0.985	0	.
3						
tenure	.9887492	.0088985	-1.26	0.209	.9714614	1.006345
exper	.9684241	.0096896	-3.21	0.001	.9496178	.9876028
pyperstudlep	1.020073	.0134123	1.51	0.131	.9941217	1.046703
pyperstude~s	1.030962	.0043982	7.15	0.000	1.022378	1.039619
pyperstudw~e	1.036865	.0205199	1.83	0.067	.9974167	1.077873
pyperstudb~k	1.002662	.0206478	0.13	0.897	.9629987	1.043959
pyperstudl~o	.987108	.0200036	-0.64	0.522	.94867	1.027103
pyperstudm~e	.9964854	.0054801	-0.64	0.522	.9858023	1.007284
male	1.527867	.1993507	3.25	0.001	1.183106	1.973091
age	1.062613	.0096426	6.69	0.000	1.043881	1.081681
Asian	.8150737	.9268982	-0.18	0.857	.0877433	7.571459
Black	2.840248	.847936	3.50	0.000	1.582108	5.0989
Latino	2.709875	.9954034	2.71	0.007	1.319113	5.566938
Other	.5059867	.5736236	-0.60	0.548	.0548472	4.667921
AsianMale	(omitted)					
BlackMale	.5247994	.2498638	-1.35	0.176	.2064074	1.334324
LatinoMale	.5952181	.3186329	-0.97	0.332	.2084538	1.699583
OtherMale	.8230852	1.381834	-0.12	0.908	.0306484	22.10453

```

-----> region = 12

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -774.53178
Iteration 1: log likelihood = -714.00865
Iteration 2: log likelihood = -711.77751
Iteration 3: log likelihood = -711.67735
Iteration 4: log likelihood = -711.66293
Iteration 5: log likelihood = -711.66126

```

```

Iteration 6:  log likelihood = -711.66092
Iteration 7:  log likelihood = -711.66085
Iteration 8:  log likelihood = -711.66083

```

```

Multinomial logistic regression      Number of obs   =      753
                                     LR chi2(32)        =     125.74
                                     Prob > chi2         =     0.0000
Log likelihood = -711.66083          Pseudo R2        =     0.0812

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	.9982037	.0157033	-0.11	0.909	.9678955	1.029461
	exper	.9953196	.0216493	-0.22	0.829	.9537795	1.038669
	pypersistudlep	1.023106	.0264513	0.88	0.377	.9725539	1.076285
	pypersistude~s	.9992262	.008075	-0.10	0.924	.9835242	1.015179
	pypersistudw~e	.8304908	.0453308	-3.40	0.001	.7462314	.9242641
	pypersistudb~k	.8356586	.0501691	-2.99	0.003	.7428937	.9400071
	pypersistudl~o	.8410294	.0475471	-3.06	0.002	.7528163	.939579
	pypersistudm~e	1.003862	.0083394	0.46	0.643	.987649	1.02034
	male	1.147559	.2651324	0.60	0.551	.7296473	1.804832
	age	.9671603	.0178386	-1.81	0.070	.9328217	1.002763
	Asian	(omitted)					
	Black	1.272525	.5316491	0.58	0.564	.5611028	2.885961
	Latino	2.664955	2.120101	1.23	0.218	.5604252	12.6725
	Other	3.92e-06	.0029024	-0.02	0.987	0	.
	AsianMale	(omitted)					
	BlackMale	2.6293	1.638469	1.55	0.121	.775188	8.918118
	LatinoMale	.9498029	1.179618	-0.04	0.967	.0832673	10.83409
	OtherMale	.2518175	435.4811	-0.00	0.999	0	.
3							
	tenure	1.016307	.0126011	1.30	0.192	.9919066	1.041307
	exper	1.048678	.0200029	2.49	0.013	1.010197	1.088625
	pypersistudlep	1.010421	.0230402	0.45	0.649	.9662569	1.056603
	pypersistude~s	1.029469	.0071107	4.20	0.000	1.015626	1.043501
	pypersistudw~e	.9409725	.0534048	-1.07	0.284	.8419128	1.051688
	pypersistudb~k	.9171448	.0552073	-1.44	0.151	.8150796	1.031991
	pypersistudl~o	.927749	.0538263	-1.29	0.196	.8280286	1.039479
	pypersistudm~e	.988409	.0093288	-1.24	0.217	.9702929	1.006863
	male	1.941811	.3793307	3.40	0.001	1.324115	2.847661
	age	.975551	.0163584	-1.48	0.140	.9440102	1.008146
	Asian	(omitted)					
	Black	1.537218	.633991	1.04	0.297	.684978	3.449804
	Latino	1.721067	1.619867	0.58	0.564	.2720493	10.888
	Other	1.681656	2.259707	0.39	0.699	.120766	23.41692
	AsianMale	(omitted)					
	BlackMale	.9032034	.5621021	-0.16	0.870	.2667144	3.058614
	LatinoMale	1.527573	1.971637	0.33	0.743	.1217187	19.17109
	OtherMale	9.60e-07	.0011966	-0.01	0.991	0	.

```

-----> region = 13

```

```

note: AsianMale omitted because of collinearity
Iteration 0:  log likelihood = -1171.4552
Iteration 1:  log likelihood = -1100.0925
Iteration 2:  log likelihood = -1098.406
Iteration 3:  log likelihood = -1098.2171
Iteration 4:  log likelihood = -1098.1857
Iteration 5:  log likelihood = -1098.1787
Iteration 6:  log likelihood = -1098.177
Iteration 7:  log likelihood = -1098.1766

```



Iteration 8: log likelihood = -1098.1765  
 Iteration 9: log likelihood = -1098.1765

Multinomial logistic regression	Number of obs	=	1087
	LR chi2(34)	=	146.56
	Prob > chi2	=	0.0000
Log likelihood = -1098.1765	Pseudo R2	=	0.0626

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.028591	.012249	2.37	0.018	1.004861	1.052881
	exper	.992192	.0159719	-0.49	0.626	.9613764	1.023995
pyperstudlep		.9722848	.016232	-1.68	0.092	.9409855	1.004625
pyperstude~s		.9930732	.0071088	-0.97	0.332	.9792375	1.007104
pyperstudw~e		.9502763	.0305956	-1.58	0.113	.8921628	1.012175
pyperstudb~k		.9755427	.0333725	-0.72	0.469	.9122785	1.043194
pyperstudl~o		.9678923	.0301889	-1.05	0.295	.9104954	1.028907
pyperstudm~e		.9855664	.0075979	-1.89	0.059	.9707867	1.000571
	male	1.290501	.2266939	1.45	0.147	.9146044	1.820889
	age	.9685326	.0139157	-2.23	0.026	.9416387	.9961946
	Asian	2.41e-07	.0004794	-0.01	0.994	0	.
	Black	1.883032	.6684434	1.78	0.075	.9390656	3.77589
	Latino	1.22333	.3619708	0.68	0.496	.6849862	2.184768
	Other	4.88e-07	.0004933	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	1.587106	1.04887	0.70	0.485	.4345829	5.796142
	LatinoMale	2.657005	1.253405	2.07	0.038	1.054023	6.697839
	OtherMale	2745403	2.77e+09	0.01	0.988	0	.
3							
	tenure	.9962357	.011325	-0.33	0.740	.9742845	1.018682
	exper	1.081152	.0185794	4.54	0.000	1.045343	1.118187
pyperstudlep		.9729647	.0167946	-1.59	0.112	.9405984	1.006445
pyperstude~s		1.026825	.0074773	3.64	0.000	1.012274	1.041585
pyperstudw~e		.9915365	.0339959	-0.25	0.804	.9270952	1.060457
pyperstudb~k		.9805413	.036152	-0.53	0.594	.9121842	1.054021
pyperstudl~o		.9824805	.0329306	-0.53	0.598	.920012	1.049191
pyperstudm~e		.9786688	.0091024	-2.32	0.020	.96099	.9966728
	male	1.365481	.2341383	1.82	0.069	.9757317	1.910912
	age	.9634969	.0150476	-2.38	0.017	.934451	.9934456
	Asian	9.31e-07	.0018871	-0.01	0.995	0	.
	Black	.9028162	.4078984	-0.23	0.821	.3724095	2.188658
	Latino	.8664457	.2757741	-0.45	0.652	.4643219	1.616827
	Other	3.61e-07	.0004211	-0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	4.339861	3.120767	2.04	0.041	1.060187	17.76517
	LatinoMale	1.628333	.8667549	0.92	0.360	.5736566	4.622046
	OtherMale	6.546375	10684.72	0.00	0.999	0	.

-----> region = 14

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -279.48608  
 Iteration 1: log likelihood = -244.66801  
 Iteration 2: log likelihood = -241.91258  
 Iteration 3: log likelihood = -241.70434  
 Iteration 4: log likelihood = -241.66036  
 Iteration 5: log likelihood = -241.64964  
 Iteration 6: log likelihood = -241.64742  
 Iteration 7: log likelihood = -241.64695

Iteration 8: log likelihood = -241.64684  
 Iteration 9: log likelihood = -241.64681  
 Iteration 10: log likelihood = -241.64681

Multinomial logistic regression      Number of obs      =      290  
    LR chi2(32)      =      75.68  
    Prob > chi2      =      0.0000  
 Log likelihood = -241.64681      Pseudo R2      =      0.1354

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	.9740233	.028796	-0.89	0.373	.9191882	1.03213
	exper	1.002037	.0487597	0.04	0.967	.910886	1.10231
	pyperstudlep	.8997001	.1209395	-0.79	0.432	.6913175	1.170895
	pyperstude~s	.9782323	.0172508	-1.25	0.212	.9449991	1.012634
	pyperstudw~e	.8390204	.1908299	-0.77	0.440	.5372446	1.310307
	pyperstudb~k	.9024042	.2300458	-0.40	0.687	.5475302	1.487285
	pyperstudl~o	.8750362	.1918461	-0.61	0.543	.5693844	1.344765
	pyperstudm~e	.9758324	.0330043	-0.72	0.469	.9132427	1.042712
	male	1.69253	.751389	1.19	0.236	.7090068	4.040382
	age	.9865313	.0438606	-0.31	0.760	.904205	1.076353
	Asian	(omitted)					
	Black	6.82e-07	.0021647	-0.00	0.996	0	.
	Latino	3.992052	5.953731	0.93	0.353	.2146432	74.24634
	Other	1.54e-06	.0047705	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	1249321	7.62e+09	0.00	0.998	0	.
	LatinoMale	.7072321	1.281595	-0.19	0.848	.0202804	24.66314
	OtherMale	.6858386	3007.744	-0.00	1.000	0	.
3							
	tenure	.9804444	.0187112	-1.03	0.301	.9444485	1.017812
	exper	1.04835	.0340095	1.46	0.146	.9837677	1.117173
	pyperstudlep	1.211751	.0899809	2.59	0.010	1.047625	1.40159
	pyperstude~s	1.018269	.0111597	1.65	0.099	.9966301	1.040379
	pyperstudw~e	1.476809	.2614788	2.20	0.028	1.043791	2.089465
	pyperstudb~k	1.445419	.2867703	1.86	0.063	.9797511	2.132414
	pyperstudl~o	1.430433	.2487793	2.06	0.040	1.017249	2.011442
	pyperstudm~e	1.018139	.0213342	0.86	0.391	.9771721	1.060824
	male	2.818055	.8579663	3.40	0.001	1.551666	5.118004
	age	.9737285	.0300623	-0.86	0.389	.9165548	1.034469
	Asian	(omitted)					
	Black	2.823268	4.740545	0.62	0.536	.1050754	75.85832
	Latino	1.36e-07	.0002246	-0.01	0.992	0	.
	Other	.9501805	1.447229	-0.03	0.973	.0480101	18.80526
	AsianMale	(omitted)					
	BlackMale	3224497	7.27e+09	0.01	0.995	0	.
	LatinoMale	5740876	9.46e+09	0.01	0.992	0	.
	OtherMale	.9663667	2.102175	-0.02	0.987	.0135985	68.67416

-----> region = 15

note: Other omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: BlackMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -303.11913  
 Iteration 1: log likelihood = -264.80503  
 Iteration 2: log likelihood = -262.59677  
 Iteration 3: log likelihood = -262.53101  
 Iteration 4: log likelihood = -262.51756

```

Iteration 5: log likelihood = -262.51432
Iteration 6: log likelihood = -262.51363
Iteration 7: log likelihood = -262.51349
Iteration 8: log likelihood = -262.51345
Iteration 9: log likelihood = -262.51345

```

```

Multinomial logistic regression
Number of obs   =      326
LR chi2(28)     =      81.21
Prob > chi2     =      0.0000
Pseudo R2      =      0.1340

Log likelihood = -262.51345

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.019142	.0302314	0.64	0.523	.9615795	1.080151
exper	1.133216	.0708896	2.00	0.046	1.002455	1.281034
pypersistudlep	.9910395	.0259406	-0.34	0.731	.9414791	1.043209
pypersistude~s	.9950358	.0125283	-0.40	0.693	.9707813	1.019896
pypersistudw~e	.7898824	.2212445	-0.84	0.400	.4561867	1.367673
pypersistudb~k	.8140474	.2512793	-0.67	0.505	.4445306	1.490726
pypersistudl~o	.8138749	.2271849	-0.74	0.461	.4709267	1.406572
pypersistudm~e	1.026981	.0190616	1.43	0.151	.9902924	1.065029
male	1.380567	.667956	0.67	0.505	.534839	3.563625
age	.8872737	.0503891	-2.11	0.035	.7938109	.9917408
Asian	8.66e-07	.0021898	-0.01	0.996	0	.
Black	4.47e-07	.0011301	-0.01	0.995	0	.
Latino	.3190362	.2907191	-1.25	0.210	.0534798	1.903226
Other	(omitted)					
AsianMale	(omitted)					
BlackMale	(omitted)					
LatinoMale	2.448793	2.504952	0.88	0.381	.329784	18.18338
OtherMale	(omitted)					
3						
tenure	.9998008	.0189948	-0.01	0.992	.9632562	1.037732
exper	1.058569	.0307473	1.96	0.050	.9999886	1.120581
pypersistudlep	.9992997	.0196067	-0.04	0.972	.9616008	1.038477
pypersistude~s	1.041141	.0092271	4.55	0.000	1.023213	1.059384
pypersistudw~e	1.009179	.1730433	0.05	0.958	.721129	1.412287
pypersistudb~k	.9489917	.1733808	-0.29	0.774	.6633579	1.357616
pypersistudl~o	.9795658	.1673849	-0.12	0.904	.7007824	1.369254
pypersistudm~e	1.012393	.0151381	0.82	0.410	.9831535	1.042502
male	2.751693	.8585839	3.24	0.001	1.492821	5.07215
age	.9784009	.0256979	-0.83	0.406	.9293084	1.030087
Asian	6.62e-06	.0089636	-0.01	0.993	0	.
Black	3.37e-06	.0045591	-0.01	0.993	0	.
Latino	.5552148	.3670441	-0.89	0.373	.1519653	2.028512
Other	(omitted)					
AsianMale	(omitted)					
BlackMale	(omitted)					
LatinoMale	2.027378	1.52872	0.94	0.349	.4624837	8.887366
OtherMale	(omitted)					

```

-----> region = 16

```

```

note: AsianMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -418.07322
Iteration 1: log likelihood = -375.77776
Iteration 2: log likelihood = -373.97576
Iteration 3: log likelihood = -370.59624
Iteration 4: log likelihood = -370.40905
Iteration 5: log likelihood = -370.36469

```

```

Iteration 6: log likelihood = -370.35475
Iteration 7: log likelihood = -370.35241
Iteration 8: log likelihood = -370.35192
Iteration 9: log likelihood = -370.35184
Iteration 10: log likelihood = -370.35183

```

```

Multinomial logistic regression
Number of obs   =      402
LR chi2(34)     =      95.44
Prob > chi2     =      0.0000
Pseudo R2      =      0.1141

Log likelihood = -370.35183

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.088206	.0307227	2.99	0.003	1.029626	1.150118
	exper	.9462429	.0329046	-1.59	0.112	.8838999	1.012983
pypersistudlep		.9688951	.0251967	-1.22	0.224	.9207479	1.01956
pypersistude~s		.9792995	.0096078	-2.13	0.033	.9606485	.9983126
pypersistudw~e		.9764366	.0335717	-0.69	0.488	.9128053	1.044503
pypersistudb~k		.9938328	.0426834	-0.14	0.885	.9135991	1.081113
pypersistudl~o		1.006481	.0337966	0.19	0.847	.9423732	1.074949
pypersistudm~e		.998812	.0158971	-0.07	0.940	.9681353	1.030461
	male	1.387883	.4605986	0.99	0.323	.7242086	2.659759
	age	.9612197	.0277672	-1.37	0.171	.908309	1.017213
	Asian	1.52e+08	2.00e+12	0.00	0.999	0	.
	Black	.6360553	.626592	-0.46	0.646	.0922482	4.385631
	Latino	1.96515	1.63004	0.81	0.415	.3866752	9.987228
	Other	1.12e-07	.0003941	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	.7463732	1.187717	-0.18	0.854	.0329932	16.88449
	LatinoMale	2.953867	3.633539	0.88	0.379	.2650522	32.91929
	OtherMale	1.443569	7660.23	0.00	1.000	0	.
3							
	tenure	1.041565	.0180084	2.36	0.019	1.006861	1.077466
	exper	1.011056	.0231667	0.48	0.631	.9666548	1.057497
pypersistudlep		1.015014	.0185192	0.82	0.414	.9793579	1.051967
pypersistude~s		1.017012	.0070166	2.44	0.014	1.003352	1.030857
pypersistudw~e		1.037022	.0387273	0.97	0.330	.9638292	1.115773
pypersistudb~k		1.063951	.0482732	1.37	0.172	.973422	1.162899
pypersistudl~o		1.022728	.0376921	0.61	0.542	.9514583	1.099337
pypersistudm~e		.9954603	.0129846	-0.35	0.727	.9703335	1.021238
	male	1.465532	.3654411	1.53	0.125	.8989635	2.389179
	age	.9996491	.022226	-0.02	0.987	.9570225	1.044174
	Asian	3.208747	61394.88	0.00	1.000	0	.
	Black	1.04e-07	.0001219	-0.01	0.989	0	.
	Latino	.6280363	.5920785	-0.49	0.622	.0989733	3.985211
	Other	1.16e-07	.0003161	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	.8833065	1789.543	-0.00	1.000	0	.
	LatinoMale	3.33e-08	.0001579	-0.00	0.997	0	.
	OtherMale	.3348154	1245.304	-0.00	1.000	0	.

```

-----> region = 17

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -419.82318
Iteration 1: log likelihood = -379.73463
Iteration 2: log likelihood = -376.97934
Iteration 3: log likelihood = -376.72435
Iteration 4: log likelihood = -376.66322

```

```

Iteration 5: log likelihood = -376.65169
Iteration 6: log likelihood = -376.64972
Iteration 7: log likelihood = -376.64926
Iteration 8: log likelihood = -376.64915
Iteration 9: log likelihood = -376.64913

```

```

Multinomial logistic regression
Number of obs   =      419
LR chi2(32)     =      86.35
Prob > chi2     =      0.0000
Pseudo R2      =      0.1028

Log likelihood = -376.64913

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9456877	.0237495	-2.22	0.026	.9002666	.9934004
exper	1.055532	.0384032	1.49	0.137	.9828842	1.13355
pyperstudlep	.9937187	.0458544	-0.14	0.891	.90779	1.087781
pyperstude~s	.9945366	.0131162	-0.42	0.678	.9691587	1.020579
pyperstudw~e	.6342248	.1042318	-2.77	0.006	.4595707	.875254
pyperstudb~k	.652666	.1073642	-2.59	0.009	.4727891	.9009786
pyperstudl~o	.6297899	.1033155	-2.82	0.005	.4566234	.8686267
pyperstudm~e	1.008922	.0120985	0.74	0.459	.9854861	1.032916
male	2.315975	.8755742	2.22	0.026	1.1039	4.8589
age	.981093	.0322124	-0.58	0.561	.9199464	1.046304
Asian	(omitted)					
Black	1.17e-06	.0014939	-0.01	0.991	0	.
Latino	.84451	1.015843	-0.14	0.888	.0799305	8.922713
Other	4.690592	23944.43	0.00	1.000	0	.
AsianMale	(omitted)					
BlackMale	3481476	4.44e+09	0.01	0.991	0	.
LatinoMale	2.996189	4.152967	0.79	0.429	.198031	45.33205
OtherMale	.1750277	1027.751	-0.00	1.000	0	.
3						
tenure	.9814091	.0164895	-1.12	0.264	.9496167	1.014266
exper	1.062367	.0281614	2.28	0.022	1.008581	1.119021
pyperstudlep	1.031257	.0252518	1.26	0.209	.9829327	1.081956
pyperstude~s	1.016037	.0102478	1.58	0.115	.996149	1.036322
pyperstudw~e	.9955523	.1490939	-0.03	0.976	.7423149	1.335181
pyperstudb~k	.9941918	.1494776	-0.04	0.969	.7404425	1.334901
pyperstudl~o	.9891345	.1472852	-0.07	0.942	.7387692	1.324347
pyperstudm~e	.9773519	.0121884	-1.84	0.066	.9537526	1.001535
male	2.241085	.5667422	3.19	0.001	1.365212	3.678887
age	.9882251	.0236254	-0.50	0.620	.9429883	1.035632
Asian	(omitted)					
Black	3.623239	3.52182	1.32	0.185	.5391648	24.34851
Latino	.8517909	.6147464	-0.22	0.824	.2070198	3.504726
Other	1.10e+07	2.77e+10	0.01	0.995	0	.
AsianMale	(omitted)					
BlackMale	.2833399	.4875981	-0.73	0.464	.0097158	8.262961
LatinoMale	2.396282	2.195574	0.95	0.340	.3977695	14.43592
OtherMale	.6409003	1868.876	-0.00	1.000	0	.

```
-----> region = 18
```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -412.27681
Iteration 1: log likelihood = -370.92401
Iteration 2: log likelihood = -368.97098
Iteration 3: log likelihood = -368.78665
Iteration 4: log likelihood = -368.7519

```

```

Iteration 5:  log likelihood = -368.7445
Iteration 6:  log likelihood = -368.74284
Iteration 7:  log likelihood = -368.74243
Iteration 8:  log likelihood = -368.74235
Iteration 9:  log likelihood = -368.74233

```

```

Multinomial logistic regression      Number of obs   =      409
                                     LR chi2(32)        =      87.07
                                     Prob > chi2         =      0.0000
Log likelihood = -368.74233          Pseudo R2       =      0.1056

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9959411	.0244772	-0.17	0.869	.9491038	1.04509
exper	.9742495	.0349448	-0.73	0.467	.908111	1.045205
pypersistudlep	.9753352	.0166973	-1.46	0.145	.943152	1.008617
pypersistude~s	1.028345	.0138576	2.07	0.038	1.00154	1.055867
pypersistudw~e	.9752117	.1825376	-0.13	0.893	.6757301	1.407423
pypersistudb~k	.9870827	.1977499	-0.06	0.948	.6665393	1.461778
pypersistudl~o	.9667311	.1816354	-0.18	0.857	.6689238	1.397123
pypersistudm~e	1.035011	.0152087	2.34	0.019	1.005627	1.065252
male	1.506889	.5551646	1.11	0.266	.7319536	3.102266
age	.9662572	.0307872	-1.08	0.281	.9077609	1.028523
Asian	(omitted)					
Black	3.269885	4.821494	0.80	0.422	.1817286	58.83581
Latino	2.002683	1.297224	1.07	0.284	.562668	7.128074
Other	1.01e-06	.0006365	-0.02	0.983	0	.
AsianMale	(omitted)					
BlackMale	1.01e-06	.0007089	-0.02	0.984	0	.
LatinoMale	1.787763	1.407244	0.74	0.460	.3821942	8.362498
OtherMale	4.10e+13	1.96e+17	0.01	0.995	0	.
3						
tenure	.9846617	.0156793	-0.97	0.332	.9544053	1.015877
exper	1.101202	.0317725	3.34	0.001	1.040657	1.165269
pypersistudlep	.9732179	.0133906	-1.97	0.048	.9473235	.9998202
pypersistude~s	1.05025	.0117535	4.38	0.000	1.027464	1.073541
pypersistudw~e	1.089106	.1801602	0.52	0.606	.7875243	1.506177
pypersistudb~k	1.050944	.1853929	0.28	0.778	.7437427	1.485036
pypersistudl~o	1.060848	.1765426	0.35	0.723	.7655955	1.469966
pypersistudm~e	.9826596	.0165075	-1.04	0.298	.9508323	1.015552
male	1.35594	.3692411	1.12	0.263	.7951456	2.312248
age	.9578168	.0254876	-1.62	0.105	.9091424	1.009097
Asian	(omitted)					
Black	2.120477	2.677172	0.60	0.552	.1785516	25.18277
Latino	1.627977	.8533313	0.93	0.353	.5827423	4.547993
Other	2.822066	3.038534	0.96	0.335	.3420375	23.28416
AsianMale	(omitted)					
BlackMale	.9799193	1.592179	-0.01	0.990	.0405651	23.67164
LatinoMale	.7757843	.523357	-0.38	0.707	.206779	2.910554
OtherMale	1.012259	7462.94	0.00	1.000	0	.

```

-----> region = 19

```

```

Iteration 0:  log likelihood = -723.09524
Iteration 1:  log likelihood = -654.28046
Iteration 2:  log likelihood = -651.46108
Iteration 3:  log likelihood = -651.36115
Iteration 4:  log likelihood = -651.3392
Iteration 5:  log likelihood = -651.33391
Iteration 6:  log likelihood = -651.33286

```

Iteration 7: log likelihood = -651.33267  
 Iteration 8: log likelihood = -651.33265

Multinomial logistic regression	Number of obs	=	673
	LR chi2(36)	=	143.53
	Prob > chi2	=	0.0000
Log likelihood = -651.33265	Pseudo R2	=	0.0992

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.036504	.0146326	2.54	0.011	1.008218	1.065584
	exper	1.048084	.0209561	2.35	0.019	1.007805	1.089973
pyperstudlep		1.015125	.0067075	2.27	0.023	1.002063	1.028357
pyperstude~s		1.004585	.0082666	0.56	0.578	.9885131	1.020919
pyperstudw~e		.8492294	.1175237	-1.18	0.238	.6474831	1.113837
pyperstudb~k		.8246419	.1196824	-1.33	0.184	.6204808	1.09598
pyperstudl~o		.8405426	.11291	-1.29	0.196	.6459784	1.093708
pyperstudm~e		1.008333	.0085374	0.98	0.327	.9917381	1.025205
	male	1.1224	.3074963	0.42	0.673	.6560694	1.920196
	age	.9604666	.0162352	-2.39	0.017	.9291676	.9928199
	Asian	2997468	4.49e+09	0.01	0.992	0	.
	Black	.5326736	.3850461	-0.87	0.384	.1291711	2.19663
	Latino	1.290617	.3445776	0.96	0.339	.7647822	2.177995
	Other	1.83e-06	.0024978	-0.01	0.992	0	.
	AsianMale	6.02e-07	.0014138	-0.01	0.995	0	.
	BlackMale	1.098661	1.250309	0.08	0.934	.1180787	10.22248
	LatinoMale	1.333177	.5096503	0.75	0.452	.6302148	2.820248
	OtherMale	3249708	1.22e+10	0.00	0.997	0	.
3							
	tenure	1.049382	.0159259	3.18	0.001	1.018628	1.081065
	exper	1.148935	.0311701	5.12	0.000	1.089438	1.211681
pyperstudlep		1.006723	.0076196	0.89	0.376	.9918987	1.021768
pyperstude~s		1.036362	.0108888	3.40	0.001	1.015239	1.057925
pyperstudw~e		.979023	.1494401	-0.14	0.890	.7258781	1.32045
pyperstudb~k		.9083508	.1456277	-0.60	0.549	.6634193	1.24371
pyperstudl~o		.9345526	.1387809	-0.46	0.649	.6985548	1.250279
pyperstudm~e		1.00478	.0108105	0.44	0.658	.9838132	1.026193
	male	.7890777	.2564099	-0.73	0.466	.4173715	1.491821
	age	.9264024	.0219751	-3.22	0.001	.8843178	.9704898
	Asian	1.037476	2569.106	0.00	1.000	0	.
	Black	.1515132	.1701895	-1.68	0.093	.016762	1.369542
	Latino	1.084094	.3334118	0.26	0.793	.5933093	1.980857
	Other	3.73e-06	.0059547	-0.01	0.994	0	.
	AsianMale	4.46e+07	1.19e+11	0.01	0.995	0	.
	BlackMale	7.562007	11.24099	1.36	0.174	.4105052	139.3014
	LatinoMale	2.198678	.9832651	1.76	0.078	.9151621	5.282327
	OtherMale	3.11e+13	8.75e+16	0.01	0.991	0	.

-----> region = 20

Iteration 0: log likelihood = -1927.4152  
 Iteration 1: log likelihood = -1802.9933  
 Iteration 2: log likelihood = -1796.3622  
 Iteration 3: log likelihood = -1796.2469  
 Iteration 4: log likelihood = -1796.2193  
 Iteration 5: log likelihood = -1796.2137  
 Iteration 6: log likelihood = -1796.2126  
 Iteration 7: log likelihood = -1796.2123  
 Iteration 8: log likelihood = -1796.2122  
 Iteration 9: log likelihood = -1796.2122

Multinomial logistic regression

Number of obs = 1993  
 LR chi2(36) = 262.41  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0681

Log likelihood = -1796.2122

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.004313	.0090378	0.48	0.632	.9867545	1.022184
	exper	1.00956	.0126407	0.76	0.447	.9850863	1.034642
	pyperstudlep	.983784	.006886	-2.34	0.020	.9703798	.9973734
	pyperstude~s	1.00536	.0053056	1.01	0.311	.9950153	1.015813
	pyperstudw~e	.9521029	.0479109	-0.98	0.329	.8626815	1.050793
	pyperstudb~k	.9538638	.0499014	-0.90	0.367	.860906	1.056859
	pyperstudl~o	.9474969	.0474942	-1.08	0.282	.8588365	1.04531
	pyperstudm~e	.9946301	.0055853	-0.96	0.338	.9837432	1.005637
	male	1.774596	.2663196	3.82	0.000	1.322381	2.381454
	age	.9734531	.0105435	-2.48	0.013	.9530061	.9943389
	Asian	8.43e-07	.0009509	-0.01	0.990	0	.
	Black	1.621511	.4952781	1.58	0.114	.8911013	2.950618
	Latino	1.551801	.2807262	2.43	0.015	1.088556	2.212185
	Other	1.680049	2.067925	0.42	0.673	.1505226	18.75176
	AsianMale	3.16e+13	1.34e+17	0.01	0.994	0	.
	BlackMale	.6161544	.290581	-1.03	0.305	.244489	1.552815
	LatinoMale	.967844	.2443383	-0.13	0.897	.5900844	1.587437
	OtherMale	.9806292	1.510206	-0.01	0.990	.0479316	20.06262
3							
	tenure	1.036078	.0097584	3.76	0.000	1.017127	1.055381
	exper	1.101147	.0182456	5.82	0.000	1.065961	1.137495
	pyperstudlep	1.010082	.0061955	1.64	0.102	.9980123	1.022299
	pyperstude~s	1.050915	.0063963	8.16	0.000	1.038453	1.063527
	pyperstudw~e	1.094466	.0647399	1.53	0.127	.9746572	1.229002
	pyperstudb~k	1.03652	.0635617	0.58	0.559	.9191373	1.168895
	pyperstudl~o	1.036616	.0611113	0.61	0.542	.9235011	1.163586
	pyperstudm~e	1.007007	.0055382	1.27	0.204	.9962103	1.017192
	male	1.436901	.2434199	2.14	0.032	1.030928	2.002745
	age	.939649	.0133664	-4.38	0.000	.9138131	.9662153
	Asian	1.519886	1.795189	0.35	0.723	.1501164	15.38842
	Black	1.176661	.3851345	0.50	0.619	.6195034	2.234906
	Latino	1.143607	.21675	0.71	0.479	.7887638	1.658084
	Other	2.938933	3.676488	0.86	0.389	.2531456	34.12
	AsianMale	.3654843	2977.105	-0.00	1.000	0	.
	BlackMale	1.060586	.5327485	0.12	0.907	.3962551	2.838686
	LatinoMale	1.05136	.2901462	0.18	0.856	.61213	1.805758
	OtherMale	.4241828	.7322504	-0.50	0.619	.0143932	12.50114

-----> region = .

note: AsianMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -57.168138  
 Iteration 1: log likelihood = -45.192669  
 Iteration 2: log likelihood = -42.564244  
 Iteration 3: log likelihood = -41.83346  
 Iteration 4: log likelihood = -41.746353  
 Iteration 5: log likelihood = -41.728343  
 Iteration 6: log likelihood = -41.724944  
 Iteration 7: log likelihood = -41.724202  
 Iteration 8: log likelihood = -41.724025  
 Iteration 9: log likelihood = -41.723986



Iteration 10: log likelihood = -41.723978  
 Iteration 11: log likelihood = -41.723976

Multinomial logistic regression	Number of obs	=	66
	LR chi2(32)	=	30.89
	Prob > chi2	=	0.5227
Log likelihood = -41.723976	Pseudo R2	=	0.2702

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.8612892	.0776687	-1.66	0.098	.7217552	1.027799
	exper	1.180217	.1584944	1.23	0.217	.9070929	1.535578
pyperstudlep		.8650638	.1172361	-1.07	0.285	.6632706	1.128251
pyperstude~s		1.016918	.0533139	0.32	0.749	.9176135	1.126968
pyperstudw~e		.6015824	.2273304	-1.34	0.179	.286838	1.261692
pyperstudb~k		.6037045	.2321092	-1.31	0.189	.2841573	1.282596
pyperstudl~o		.6029982	.2336492	-1.31	0.192	.2821585	1.288661
pyperstudm~e		.9984752	.0333865	-0.05	0.964	.9351369	1.066103
	male	6.964683	12.91978	1.05	0.295	.1836061	264.1896
	age	.891545	.1103981	-0.93	0.354	.6994245	1.136438
	Asian	.0009329	5.992781	-0.00	0.999	0	.
	Black	12.44361	29.53051	1.06	0.288	.1188282	1303.085
	Latino	40.93658	100.6063	1.51	0.131	.3312741	5058.662
	Other	5.56e-08	.0002488	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	10.62339	42.80998	0.59	0.558	.0039455	28603.69
	LatinoMale	.0496313	.1444623	-1.03	0.302	.0001653	14.9058
	OtherMale	(omitted)					
3							
	tenure	1.029143	.0547053	0.54	0.589	.9273192	1.142148
	exper	1.04476	.0732588	0.62	0.532	.9106047	1.198679
pyperstudlep		.9798021	.0454965	-0.44	0.660	.894568	1.073157
pyperstude~s		1.01037	.0349028	0.30	0.765	.9442262	1.081147
pyperstudw~e		1.115373	.2977904	0.41	0.683	.6609374	1.882262
pyperstudb~k		1.095516	.2952931	0.34	0.735	.645922	1.858051
pyperstudl~o		1.10487	.2943355	0.37	0.708	.6554697	1.862387
pyperstudm~e		1.009012	.0235213	0.38	0.700	.9639485	1.056182
	male	1.68902	1.433024	0.62	0.537	.3202175	8.908904
	age	.9458714	.0576578	-0.91	0.361	.8393539	1.065906
	Asian	3.53e-07	.0020692	-0.00	0.998	0	.
	Black	1.93e-07	.0003485	-0.01	0.993	0	.
	Latino	1.17882	2.33415	0.08	0.934	.0243223	57.13332
	Other	1.29e-07	.0004588	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	.2356184	1895.031	-0.00	1.000	0	.
	LatinoMale	.336281	.7311403	-0.50	0.616	.0047427	23.84405
	OtherMale	(omitted)					

```
. bysort pyschooltype : mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age
> Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1997a==1,
rrr
```

-----> pyschooltype =

Iteration 0: log likelihood = -443.4197  
 Iteration 1: log likelihood = -413.03097  
 Iteration 2: log likelihood = -412.07371  
 Iteration 3: log likelihood = -411.90407  
 Iteration 4: log likelihood = -411.88054

```

Iteration 5: log likelihood = -411.87775
Iteration 6: log likelihood = -411.87721
Iteration 7: log likelihood = -411.87709
Iteration 8: log likelihood = -411.87707
Iteration 9: log likelihood = -411.87706

```

```

Multinomial logistic regression
Number of obs   =      426
LR chi2(36)     =      63.09
Prob > chi2     =      0.0035
Pseudo R2      =      0.0711

Log likelihood = -411.87706

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9575667	.0209752	-1.98	0.048	.9173262	.9995726
exper	1.054487	.0282074	1.98	0.047	1.000626	1.111248
pyperstudlep	1.015179	.0122403	1.25	0.212	.9914696	1.039455
pyperstude~s	.9949086	.0082151	-0.62	0.536	.9789369	1.011141
pyperstudw~e	.9786776	.0324258	-0.65	0.515	.9171438	1.04434
pyperstudb~k	.9825046	.0347982	-0.50	0.618	.9166149	1.053131
pyperstudl~o	.9828116	.0325635	-0.52	0.601	.9210165	1.048753
pyperstudm~e	.9956991	.0108922	-0.39	0.694	.9745781	1.017278
male	2.604397	.8678358	2.87	0.004	1.355409	5.004309
age	.9574429	.0227283	-1.83	0.067	.9139166	1.003042
Asian	4.93e-07	.0014006	-0.01	0.996	0	.
Black	2.39816	1.294543	1.62	0.105	.8325178	6.908164
Latino	1.431619	.8342184	0.62	0.538	.4569031	4.485707
Other	4.46e-07	.0008795	-0.01	0.994	0	.
AsianMale	.8429557	3388.292	-0.00	1.000	0	.
BlackMale	.6569313	.5559627	-0.50	0.620	.1250679	3.450596
LatinoMale	.3326928	.2517455	-1.45	0.146	.0755001	1.466019
OtherMale	1.013147	2616.016	0.00	1.000	0	.
3						
tenure	.9968271	.0176077	-0.18	0.857	.9629073	1.031942
exper	1.071902	.0250003	2.98	0.003	1.024005	1.122039
pyperstudlep	1.006765	.0121361	0.56	0.576	.983257	1.030834
pyperstude~s	1.016663	.0080798	2.08	0.038	1.000949	1.032623
pyperstudw~e	1.003894	.0345781	0.11	0.910	.9383592	1.074006
pyperstudb~k	.9884411	.0367067	-0.31	0.754	.9190531	1.063068
pyperstudl~o	.9856043	.0342725	-0.42	0.677	.9206693	1.055119
pyperstudm~e	.9964643	.0115483	-0.31	0.760	.9740851	1.019358
male	2.247611	.6576712	2.77	0.006	1.266636	3.988324
age	.9790473	.0205348	-1.01	0.313	.9396159	1.020134
Asian	8.85e-07	.0020999	-0.01	0.995	0	.
Black	1.247762	.7133627	0.39	0.699	.4069001	3.826268
Latino	1.587287	.9195086	0.80	0.425	.5099856	4.940295
Other	2.96e-07	.0004963	-0.01	0.993	0	.
AsianMale	1.024268	3438.872	0.00	1.000	0	.
BlackMale	.9666466	.8638469	-0.04	0.970	.1677215	5.571174
LatinoMale	.9481033	.6256689	-0.08	0.936	.2600961	3.45603
OtherMale	2965767	4.98e+09	0.01	0.993	0	.

by psychooltype

```
-----> psychooltype = Both Elem/Sec
```

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -643.61923
Iteration 1: log likelihood = -584.72524
Iteration 2: log likelihood = -578.17985
Iteration 3: log likelihood = -578.02481
Iteration 4: log likelihood = -578.00505

```

```

Iteration 5: log likelihood = -578.0017
Iteration 6: log likelihood = -578.00093
Iteration 7: log likelihood = -578.00075
Iteration 8: log likelihood = -578.00071
Iteration 9: log likelihood = -578.0007

```

```

Multinomial logistic regression
Number of obs   =      675
LR chi2(34)     =     131.24
Prob > chi2     =     0.0000
Pseudo R2      =     0.1020

Log likelihood = -578.0007

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.027861	.021277	1.33	0.184	.9869934	1.070421
exper	.9652542	.0253738	-1.35	0.179	.9167819	1.016289
pyperstudlep	1.011565	.0154148	0.75	0.451	.9817991	1.042233
pyperstude~s	.9860826	.0074487	-1.86	0.064	.9715909	1.00079
pyperstudw~e	.9215236	.0385748	-1.95	0.051	.8489368	1.000317
pyperstudb~k	.9315877	.0401575	-1.64	0.100	.8561136	1.013716
pyperstudl~o	.9252636	.0399262	-1.80	0.072	.8502274	1.006922
pyperstudm~e	.9927702	.0050243	-1.43	0.152	.9829715	1.002667
male	1.286115	.4131801	0.78	0.433	.6852057	2.414008
age	.954503	.0220795	-2.01	0.044	.9121944	.998774
Asian	2.746615	12086.97	0.00	1.000	0	.
Black	1.71057	.9041704	1.02	0.310	.6070356	4.820228
Latino	5.117006	3.064401	2.73	0.006	1.582196	16.54899
Other	1.06e-06	.0029539	-0.00	0.996	0	.
AsianMale	(omitted)					
BlackMale	1.164212	.843666	0.21	0.834	.2813162	4.818032
LatinoMale	.8563481	.703603	-0.19	0.850	.1711102	4.285729
OtherMale	6998419	1.96e+10	0.01	0.996	0	.
3						
tenure	1.007899	.0126879	0.62	0.532	.983335	1.033076
exper	1.059803	.020822	2.96	0.003	1.019768	1.101409
pyperstudlep	1.00429	.0121255	0.35	0.723	.9808037	1.028339
pyperstude~s	1.01773	.005585	3.20	0.001	1.006842	1.028735
pyperstudw~e	.988052	.0387817	-0.31	0.759	.9148914	1.067063
pyperstudb~k	.9861104	.040007	-0.34	0.730	.9107346	1.067725
pyperstudl~o	.9763338	.0391386	-0.60	0.550	.9025596	1.056138
pyperstudm~e	.9939471	.0040403	-1.49	0.135	.9860597	1.001898
male	1.956712	.4133134	3.18	0.001	1.293388	2.960226
age	.9346984	.0172995	-3.65	0.000	.9013997	.9692272
Asian	1.00e+07	1.90e+10	0.01	0.993	0	.
Black	.3207735	.2167414	-1.68	0.092	.0853209	1.205984
Latino	1.382929	.7608341	0.59	0.556	.4704371	4.065351
Other	5.10e-07	.0008922	-0.01	0.993	0	.
AsianMale	(omitted)					
BlackMale	3.95877	3.112483	1.75	0.080	.8478604	18.48401
LatinoMale	1.61464	1.125658	0.69	0.492	.4117696	6.331364
OtherMale	2862256	5.01e+09	0.01	0.993	0	.

```

-----> pyschooltype = Elementary

```

```

Iteration 0: log likelihood = -8842.8502
Iteration 1: log likelihood = -8331.937
Iteration 2: log likelihood = -8326.2702
Iteration 3: log likelihood = -8325.9751
Iteration 4: log likelihood = -8325.9534
Iteration 5: log likelihood = -8325.9487
Iteration 6: log likelihood = -8325.9479

```

Iteration 7: log likelihood = -8325.9478  
 Iteration 8: log likelihood = -8325.9478

Multinomial logistic regression	Number of obs	=	8417
	LR chi2(36)	=	1033.80
	Prob > chi2	=	0.0000
Log likelihood = -8325.9478	Pseudo R2	=	0.0585

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.003316	.005046	0.66	0.510	.9934744	1.013255
	exper	.9940123	.0061893	-0.96	0.335	.9819552	1.006217
pyperstudlep		1.009751	.0024039	4.08	0.000	1.00505	1.014473
pyperstude~s		.9843296	.0022276	-6.98	0.000	.9799732	.9887053
pyperstudw~e		.972361	.0076298	-3.57	0.000	.9575212	.9874308
pyperstudb~k		.9770187	.0081305	-2.79	0.005	.9612126	.9930848
pyperstudl~o		.9807361	.0077841	-2.45	0.014	.9655976	.9961118
pyperstudm~e		1.014102	.0045411	3.13	0.002	1.00524	1.023041
	male	1.988728	.1859368	7.35	0.000	1.655739	2.388684
	age	.9883036	.0049838	-2.33	0.020	.9785835	.9981202
	Asian	.5997038	.391575	-0.78	0.434	.1667812	2.156386
	Black	1.364152	.1611347	2.63	0.009	1.082226	1.71952
	Latino	1.053499	.1092339	0.50	0.615	.8597572	1.290899
	Other	.900062	.3994902	-0.24	0.812	.3771107	2.148207
	AsianMale	2.16892	2.307736	0.73	0.467	.2695029	17.45516
	BlackMale	.6832132	.1505356	-1.73	0.084	.4436158	1.052217
	LatinoMale	.5659002	.1007215	-3.20	0.001	.3992446	.8021224
	OtherMale	1.10e-06	.0005584	-0.03	0.979	0	.
3							
	tenure	1.010201	.0039447	2.60	0.009	1.002499	1.017962
	exper	1.051535	.0057468	9.19	0.000	1.040332	1.062859
pyperstudlep		1.000302	.002094	0.14	0.885	.9962061	1.004415
pyperstude~s		1.003962	.0020485	1.94	0.053	.9999549	1.007985
pyperstudw~e		1.006006	.0079701	0.76	0.450	.9905055	1.021749
pyperstudb~k		.9951405	.0083258	-0.58	0.560	.9789554	1.011593
pyperstudl~o		.9980354	.0080631	-0.24	0.808	.9823564	1.013965
pyperstudm~e		1.001573	.004015	0.39	0.695	.9937348	1.009474
	male	3.125825	.2376101	14.99	0.000	2.69315	3.628014
	age	1.003986	.0046821	0.85	0.394	.9948511	1.013205
	Asian	.6316284	.3496289	-0.83	0.407	.2134476	1.869098
	Black	1.076501	.1177648	0.67	0.500	.8687534	1.333928
	Latino	1.268122	.1161814	2.59	0.010	1.059685	1.517559
	Other	.4899295	.233138	-1.50	0.134	.1927883	1.245049
	AsianMale	2.628991	2.335802	1.09	0.277	.4608022	14.99904
	BlackMale	.5169121	.1024638	-3.33	0.001	.3505009	.7623322
	LatinoMale	.4829243	.0715476	-4.91	0.000	.3612179	.6456375
	OtherMale	1.16071	.8869762	0.20	0.845	.2595757	5.190195

-----> pyschooltype = Middle

Iteration 0: log likelihood = -4940.526  
 Iteration 1: log likelihood = -4666.7796  
 Iteration 2: log likelihood = -4659.7766  
 Iteration 3: log likelihood = -4659.0644  
 Iteration 4: log likelihood = -4658.9444  
 Iteration 5: log likelihood = -4658.9322  
 Iteration 6: log likelihood = -4658.9297  
 Iteration 7: log likelihood = -4658.9292  
 Iteration 8: log likelihood = -4658.9291  
 Iteration 9: log likelihood = -4658.9291

Multinomial logistic regression

Number of obs = 4753  
 LR chi2(36) = 563.19  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0570

Log likelihood = -4658.9291

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.005224	.0056085	0.93	0.350	.9942913	1.016277
exper	1.008053	.0071762	1.13	0.260	.9940853	1.022216
pyperstudlep	1.013711	.0041532	3.32	0.001	1.005604	1.021884
pyperstude~s	.9950947	.002902	-1.69	0.092	.989423	1.000799
pyperstudw~e	.9810694	.0100228	-1.87	0.061	.9616203	1.000912
pyperstudb~k	.981946	.0106978	-1.67	0.094	.9612008	1.003139
pyperstudl~o	.9769395	.0102941	-2.21	0.027	.9569704	.9973253
pyperstudm~e	1.007764	.0043029	1.81	0.070	.9993661	1.016233
male	1.675833	.1439691	6.01	0.000	1.416136	1.983155
age	.9651465	.0058674	-5.84	0.000	.9537149	.9767152
Asian	.4569259	.497473	-0.72	0.472	.0540891	3.859953
Black	1.085561	.1605309	0.56	0.579	.8124186	1.450536
Latino	2.060593	.3080963	4.84	0.000	1.53717	2.762246
Other	5.34e-07	.0003124	-0.02	0.980	0	.
AsianMale	1.48e-06	.0009054	-0.02	0.982	0	.
BlackMale	1.871612	.3813477	3.08	0.002	1.255395	2.790302
LatinoMale	1.062816	.1992888	0.32	0.745	.7359512	1.534854
OtherMale	2004501	1.17e+09	0.02	0.980	0	.
3						
tenure	1.006079	.005646	1.08	0.280	.9950733	1.017206
exper	1.06582	.009214	7.37	0.000	1.047913	1.084033
pyperstudlep	1.002321	.0049435	0.47	0.638	.9926789	1.012057
pyperstude~s	1.012139	.0034508	3.54	0.000	1.005399	1.018925
pyperstudw~e	1.029222	.0140644	2.11	0.035	1.002022	1.05716
pyperstudb~k	1.010331	.0146967	0.71	0.480	.9819329	1.039551
pyperstudl~o	1.009842	.0142419	0.69	0.487	.9823107	1.038145
pyperstudm~e	.9972093	.005691	-0.49	0.624	.9861172	1.008426
male	2.423853	.226684	9.47	0.000	2.017902	2.911471
age	.961439	.0074709	-5.06	0.000	.9469073	.9761937
Asian	6.84e-07	.0006162	-0.02	0.987	0	.
Black	.9173675	.1835654	-0.43	0.666	.6197518	1.357904
Latino	1.966835	.3593748	3.70	0.000	1.374797	2.813826
Other	.8381762	.6545404	-0.23	0.821	.1813941	3.872999
AsianMale	2343095	2.11e+09	0.02	0.987	0	.
BlackMale	1.583132	.4071385	1.79	0.074	.956338	2.620735
LatinoMale	.8287868	.1818048	-0.86	0.392	.5391642	1.273986
OtherMale	1.925949	1.928865	0.65	0.513	.270492	13.71308

-----> pyschooltype = Secondary

Iteration 0: log likelihood = -5821.4525  
 Iteration 1: log likelihood = -5541.7681  
 Iteration 2: log likelihood = -5528.5892  
 Iteration 3: log likelihood = -5528.413  
 Iteration 4: log likelihood = -5528.3746  
 Iteration 5: log likelihood = -5528.3654  
 Iteration 6: log likelihood = -5528.3634  
 Iteration 7: log likelihood = -5528.363  
 Iteration 8: log likelihood = -5528.3629  
 Iteration 9: log likelihood = -5528.3629

Multinomial logistic regression

Number of obs = 6170

```

Log likelihood = -5528.3629
LR chi2(36)      =      586.18
Prob > chi2      =      0.0000
Pseudo R2       =      0.0503

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9975777	.0044785	-0.54	0.589	.9888385	1.006394
	exper	1.001458	.0061909	0.24	0.814	.9893973	1.013666
pyperstudlep		1.013075	.004344	3.03	0.002	1.004597	1.021625
pyperstude~s		.9968904	.0024112	-1.29	0.198	.9921758	1.001627
pyperstudw~e		.9818739	.0078678	-2.28	0.022	.9665736	.9974163
pyperstudb~k		.9810071	.0082494	-2.28	0.023	.9649711	.9973096
pyperstudl~o		.9809111	.0080278	-2.36	0.019	.9653024	.9967722
pyperstudm~e		1.000887	.0026823	0.33	0.741	.9956438	1.006158
	male	1.237788	.0934297	2.83	0.005	1.067571	1.435146
	age	.979089	.0055545	-3.73	0.000	.9682626	.9900364
	Asian	1.290012	.8996754	0.37	0.715	.3288153	5.060991
	Black	1.63139	.2360202	3.38	0.001	1.228601	2.166231
	Latino	1.332059	.1937558	1.97	0.049	1.001639	1.771477
	Other	.3509014	.2638348	-1.39	0.164	.080387	1.531737
	AsianMale	2.310087	2.188732	0.88	0.377	.3606972	14.79496
	BlackMale	1.552389	.2941777	2.32	0.020	1.070774	2.250626
	LatinoMale	1.779456	.3070908	3.34	0.001	1.268793	2.495652
	OtherMale	3.102742	2.732274	1.29	0.199	.5523035	17.43065
3							
	tenure	.9796612	.0049918	-4.03	0.000	.9699261	.989494
	exper	1.060918	.0088079	7.12	0.000	1.043795	1.078323
pyperstudlep		.9832751	.0061818	-2.68	0.007	.9712334	.9954661
pyperstude~s		1.022816	.0030761	7.50	0.000	1.016805	1.028863
pyperstudw~e		1.053981	.0154412	3.59	0.000	1.024147	1.084684
pyperstudb~k		1.035141	.0157804	2.27	0.023	1.00467	1.066537
pyperstudl~o		1.035629	.0154436	2.35	0.019	1.005799	1.066345
pyperstudm~e		1.014122	.0028181	5.05	0.000	1.008614	1.019661
	male	2.41282	.2425478	8.76	0.000	1.981336	2.938269
	age	.9705759	.0078065	-3.71	0.000	.9553954	.9859976
	Asian	5.40e-06	.0026904	-0.02	0.981	0	.
	Black	1.095828	.2788899	0.36	0.719	.6654418	1.804573
	Latino	1.266341	.2766015	1.08	0.280	.8253277	1.94301
	Other	.8225142	.6398789	-0.25	0.802	.1790387	3.778677
	AsianMale	557006.3	2.77e+08	0.03	0.979	0	.
	BlackMale	1.077386	.3357879	0.24	0.811	.5848956	1.984561
	LatinoMale	1.294367	.3179795	1.05	0.294	.7997385	2.094919
	OtherMale	1.379918	1.293593	0.34	0.731	.2197362	8.665731

### 1997-98

```

. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino Oth
> er AsianMale BlackMale LatinoMale OtherMale if validcert1998a==1, rrr

```

```

Iteration 0: log likelihood = -21695.433
Iteration 1: log likelihood = -20705.42
Iteration 2: log likelihood = -20686.782
Iteration 3: log likelihood = -20686.751
Iteration 4: log likelihood = -20686.751

```

```

Multinomial logistic regression
Log likelihood = -20686.751
Number of obs      =      20775
LR chi2(36)       =      2017.36
Prob > chi2       =      0.0000
Pseudo R2        =      0.0465

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2	tenure	.9995903	.0027129	-0.15	0.880	.9942872	1.004922
	exper	.9958735	.0034401	-1.20	0.231	.9891537	1.002639
	pyperstudlep	1.008212	.001548	5.33	0.000	1.005182	1.011251
	pyperstude~s	.9939145	.0012087	-5.02	0.000	.9915483	.9962863
	pyperstudw~e	.9738124	.0043673	-5.92	0.000	.9652902	.9824099
	pyperstudb~k	.9762431	.0046423	-5.06	0.000	.9671867	.9853843
	pyperstudl~o	.9754077	.004458	-5.45	0.000	.9667092	.9841846
	pyperstudm~e	.9986165	.0015477	-0.89	0.372	.9955876	1.001655
	male	1.282387	.0558176	5.71	0.000	1.177523	1.39659
	age	.9805898	.0029317	-6.56	0.000	.9748606	.9863527
	Asian	.9474447	.3750735	-0.14	0.892	.4360949	2.058386
	Black	1.370411	.1004842	4.30	0.000	1.186964	1.582212
	Latino	1.407334	.0968508	4.97	0.000	1.229756	1.610556
	Other	.678303	.2063081	-1.28	0.202	.3737042	1.231174
	AsianMale	1.546292	.8880739	0.76	0.448	.5016783	4.76604
	BlackMale	1.544104	.1674902	4.01	0.000	1.248377	1.909884
	LatinoMale	1.391377	.1289505	3.56	0.000	1.160264	1.668525
	OtherMale	2.01054	.8568077	1.64	0.101	.8721003	4.635099
3	tenure	1.00635	.0024563	2.59	0.010	1.001547	1.011176
	exper	1.04787	.0037984	12.90	0.000	1.040452	1.055342
	pyperstudlep	1.008404	.0015273	5.53	0.000	1.005415	1.011402
	pyperstude~s	1.025308	.0012766	20.07	0.000	1.022809	1.027813
	pyperstudw~e	1.01175	.0053237	2.22	0.026	1.00137	1.022239
	pyperstudb~k	.9891742	.0055175	-1.95	0.051	.978419	1.000048
	pyperstudl~o	.9861519	.005309	-2.59	0.010	.9758012	.9966124
	pyperstudm~e	1.000588	.0016139	0.36	0.716	.9974295	1.003756
	male	1.512093	.0622815	10.04	0.000	1.394821	1.639226
	age	.9880894	.0032334	-3.66	0.000	.9817723	.9944472
	Asian	.4376673	.2210356	-1.64	0.102	.1626525	1.17768
	Black	.9581197	.0776964	-0.53	0.598	.8173229	1.123171
	Latino	1.320236	.0924505	3.97	0.000	1.150922	1.51446
	Other	.5169652	.1757339	-1.94	0.052	.2655273	1.006499
	AsianMale	3.647593	2.327812	2.03	0.043	1.044212	12.7416
	BlackMale	1.262075	.1543182	1.90	0.057	.9931291	1.603852
	LatinoMale	.9632647	.0926459	-0.39	0.697	.7977703	1.16309
	OtherMale	2.238674	1.023282	1.76	0.078	.9139288	5.483643

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=20775)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	7.903	2	0.019
exper	203.566	2	0.000
pyperstudlep	43.628	2	0.000
pyperstude~s	566.723	2	0.000
pyperstudw~e	54.010	2	0.000
pyperstudb~k	25.538	2	0.000
pyperstudl~o	30.312	2	0.000
pyperstudm~e	1.224	2	0.542
male	108.605	2	0.000
age	47.065	2	0.000
Asian	3.216	2	0.200
Black	21.488	2	0.000
Latino	30.931	2	0.000
Other	5.006	2	0.082

AsianMale		4.476	2	0.107
BlackMale		16.382	2	0.000
LatinoMale		16.164	2	0.000
OtherMale		4.685	2	0.096

\*\*\*\* Wald tests for independent variables (N=20775)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		7.875	2	0.019
exper		192.976	2	0.000
pyperstudlep		43.216	2	0.000
pyperstude~s		535.324	2	0.000
pyperstudw~e		54.975	2	0.000
pyperstudb~k		25.650	2	0.000
pyperstudl~o		30.346	2	0.000
pyperstudm~e		1.214	2	0.545
male		108.242	2	0.000
age		46.563	2	0.000
Asian		2.713	2	0.258
Black		21.906	2	0.000
Latino		30.936	2	0.000
Other		4.660	2	0.097
AsianMale		4.150	2	0.126
BlackMale		16.289	2	0.000
LatinoMale		16.152	2	0.000
OtherMale		4.572	2	0.102

\*\*\*\* Hausman tests of IIA assumption (N=20775)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-82.828	18	---	---
3		-5.552	18	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=20775)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-4768.330	-4759.499	17.663	19	0.545	for Ho
3		-4772.761	-4762.482	20.558	19	0.362	for Ho

\*\*\*\* Wald tests for combining alternatives (N=20775)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-		1300.011	18	0.000
2-		501.925	18	0.000
3-		1086.189	18	0.000

\*\*\*\* LR tests for combining alternatives (N=20775)



Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1454.013	18	0.000
2-	1	514.145	18	0.000
3-	1	1201.506	18	0.000

by region

```
. bysort region: mlogit admin tenure exper pyperstudlep pyperstuddecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1998a==1, rrr
```

```
-----> region = 1
```

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -1437.8992
Iteration 1: log likelihood = -1334.7904
Iteration 2: log likelihood = -1331.7568
Iteration 3: log likelihood = -1331.5694
Iteration 4: log likelihood = -1331.5367
Iteration 5: log likelihood = -1331.5335
Iteration 6: log likelihood = -1331.5328
Iteration 7: log likelihood = -1331.5327
Iteration 8: log likelihood = -1331.5327
Iteration 9: log likelihood = -1331.5326
```

Multinomial logistic regression	Number of obs	=	1347
	LR chi2(34)	=	212.73
	Prob > chi2	=	0.0000
Log likelihood = -1331.5326	Pseudo R2	=	0.0740

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.007365	.0103485	0.71	0.475	.987285 1.027853
exper	1.031215	.0146815	2.16	0.031	1.002837 1.060395
pyperstudlep	.9988833	.0038064	-0.29	0.769	.9914506 1.006372
pyperstude~s	.9894091	.0071283	-1.48	0.139	.9755361 1.003479
pyperstudw~e	1.246974	.161726	1.70	0.089	.9670773 1.607881
pyperstudb~k	1.292683	.2834144	1.17	0.242	.8411438 1.986616
pyperstudl~o	1.263618	.1551127	1.91	0.057	.9934096 1.607323
pyperstudm~e	.9889006	.0082292	-1.34	0.180	.9729024 1.005162
male	1.312453	.3821433	0.93	0.350	.7417229 2.32234
age	.9549353	.0120561	-3.65	0.000	.9315957 .9788596
Asian	3.63e-07	.0007385	-0.01	0.994	0 .
Black	3.076838	4.447139	0.78	0.437	.1810557 52.28742
Latino	1.185195	.2719094	0.74	0.459	.7559719 1.85812
Other	1.968793	2.83772	0.47	0.638	.116767 33.19554
AsianMale	(omitted)				
BlackMale	9.35e-08	.0002663	-0.01	0.995	0 .
LatinoMale	1.698102	.5561998	1.62	0.106	.8936337 3.22677
OtherMale	.5241891	1.02223	-0.33	0.740	.0114695 23.95689
3					
tenure	1.034992	.0107798	3.30	0.001	1.014078 1.056337
exper	1.090755	.0195095	4.86	0.000	1.05318 1.129671
pyperstudlep	1.012264	.0040114	3.08	0.002	1.004432 1.020157
pyperstude~s	1.02125	.0088422	2.43	0.015	1.004065 1.038728
pyperstudw~e	1.125792	.1450933	0.92	0.358	.8744894 1.449312

pyperstudb~k		.8988325	.1991187	-0.48	0.630	.582253	1.387541
pyperstudl~o		1.060322	.1289353	0.48	0.630	.8354712	1.345687
pyperstudm~e		1.014384	.007043	2.06	0.040	1.000674	1.028282
male		.6678467	.2302723	-1.17	0.242	.3397693	1.312712
age		.9518792	.0154099	-3.05	0.002	.9221506	.9825662
Asian		8.83e-08	.0001795	-0.01	0.994	0	.
Black		3.49e-06	.0060865	-0.01	0.994	0	.
Latino		1.071025	.2469612	0.30	0.766	.6815945	1.682957
Other		7.53e-07	.0012959	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		.0465361	163.2408	-0.00	0.999	0	.
LatinoMale		2.450396	.934395	2.35	0.019	1.16052	5.173922
OtherMale		1215666	2.09e+09	0.01	0.994	0	.

-----> region = 2

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -825.0379  
Iteration 1: log likelihood = -771.06518  
Iteration 2: log likelihood = -769.10322  
Iteration 3: log likelihood = -769.01737  
Iteration 4: log likelihood = -768.99988  
Iteration 5: log likelihood = -768.99617  
Iteration 6: log likelihood = -768.99534  
Iteration 7: log likelihood = -768.99514  
Iteration 8: log likelihood = -768.9951  
Iteration 9: log likelihood = -768.99509

Multinomial logistic regression	Number of obs	=	864
	LR chi2(32)	=	112.09
	Prob > chi2	=	0.0000
Log likelihood = -768.99509	Pseudo R2	=	0.0679

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9993025	.0160527	-0.04	0.965	.9683299 1.031266
exper		.9642519	.0185941	-1.89	0.059	.9284882 1.001393
pyperstudlep		1.002942	.0209461	0.14	0.888	.9627177 1.044848
pyperstude~s		1.000856	.0058999	0.15	0.885	.989359 1.012487
pyperstudw~e		1.031247	.0667846	0.48	0.635	.9083179 1.170812
pyperstudb~k		1.035225	.0670764	0.53	0.593	.911763 1.175405
pyperstudl~o		1.024474	.0629576	0.39	0.694	.9082207 1.155607
pyperstudm~e		1.004122	.0087272	0.47	0.636	.9871617 1.021373
male		1.298681	.3426959	0.99	0.322	.7742614 2.178299
age		1.009037	.0157838	0.58	0.565	.9785711 1.040452
Asian		(omitted)				
Black		1.440919	1.051647	0.50	0.617	.3446632 6.023993
Latino		1.2089	.3320947	0.69	0.490	.7055998 2.071202
Other		2.91e-06	.0023769	-0.02	0.988	0 .
AsianMale		(omitted)				
BlackMale		6.66632	7.430114	1.70	0.089	.7501673 59.23988
LatinoMale		1.388019	.5463776	0.83	0.405	.6416963 3.002348
OtherMale		1043421	8.51e+08	0.02	0.986	0 .
3						
tenure		.9829448	.0121834	-1.39	0.165	.9593535 1.007116
exper		1.0792	.0202107	4.07	0.000	1.040306 1.119548
pyperstudlep		1.045281	.0166751	2.78	0.006	1.013105 1.07848
pyperstude~s		1.019336	.0057955	3.37	0.001	1.00804 1.030758
pyperstudw~e		1.200012	.0790596	2.77	0.006	1.054645 1.365415

pyperstudb~k		1.151543	.0790572	2.06	0.040	1.006567	1.317401
pyperstudl~o		1.166942	.0734981	2.45	0.014	1.031425	1.320265
pyperstudm~e		.9988253	.0092397	-0.13	0.899	.9808789	1.0171
male		1.119292	.2825925	0.45	0.655	.6823962	1.835904
age		.9855719	.0170901	-0.84	0.402	.9526387	1.019644
Asian		(omitted)					
Black		1.607519	1.100546	0.69	0.488	.4201493	6.150477
Latino		1.030705	.2602979	0.12	0.905	.6283031	1.69083
Other		2.846138	3.567636	0.83	0.404	.2439352	33.2076
AsianMale		(omitted)					
BlackMale		5.935456	6.522491	1.62	0.105	.6887571	51.14958
LatinoMale		1.846893	.6682692	1.70	0.090	.9087571	3.753493
OtherMale		1.84e-06	.0020617	-0.01	0.991	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -348.98186  
Iteration 1: log likelihood = -308.0239  
Iteration 2: log likelihood = -305.47803  
Iteration 3: log likelihood = -305.24046  
Iteration 4: log likelihood = -305.21062  
Iteration 5: log likelihood = -305.20571  
Iteration 6: log likelihood = -305.20455  
Iteration 7: log likelihood = -305.2043  
Iteration 8: log likelihood = -305.20424  
Iteration 9: log likelihood = -305.20423

Multinomial logistic regression	Number of obs	=	338
	LR chi2(30)	=	87.56
	Prob > chi2	=	0.0000
Log likelihood = -305.20423	Pseudo R2	=	0.1254

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9987241	.0236587	-0.05	0.957	.9534138 1.046188
exper		1.01242	.0361231	0.35	0.729	.944039 1.085754
pyperstudlep		.9269194	.0562502	-1.25	0.211	.8229752 1.043992
pyperstude~s		.9827815	.0137449	-1.24	0.214	.9562079 1.010094
pyperstudw~e		.9844437	.0505265	-0.31	0.760	.8902315 1.088626
pyperstudb~k		1.039118	.052358	0.76	0.446	.9414025 1.146976
pyperstudl~o		1.028463	.0534954	0.54	0.590	.9287809 1.138843
pyperstudm~e		1.001803	.0158439	0.11	0.909	.9712258 1.033343
male		1.065877	.4017295	0.17	0.866	.5092004 2.231132
age		.941899	.0295614	-1.91	0.056	.8857057 1.001657
Asian		(omitted)				
Black		2.225279	2.155632	0.83	0.409	.333291 14.85748
Latino		3.02153	1.966873	1.70	0.089	.8435971 10.82228
Other		974227.2	6.73e+08	0.02	0.984	0 .
AsianMale		(omitted)				
BlackMale		1.330009	2.183684	0.17	0.862	.053248 33.22045
LatinoMale		1.102595	1.023164	0.11	0.916	.1788691 6.796678
OtherMale		(omitted)				
3						
tenure		.9827765	.0178047	-0.96	0.338	.9484922 1.0183
exper		1.065604	.0301212	2.25	0.025	1.008173 1.126307
pyperstudlep		1.089623	.0415138	2.25	0.024	1.011222 1.174104
pyperstude~s		1.020111	.0115011	1.77	0.077	.9978163 1.042904

pyperstudw~e		1.11901	.0506427	2.48	0.013	1.024027	1.222803
pyperstudb~k		1.133615	.0504162	2.82	0.005	1.038985	1.236864
pyperstudl~o		1.106508	.0508825	2.20	0.028	1.011142	1.210868
pyperstudm~e		1.01623	.0150485	1.09	0.277	.9871591	1.046156
male		3.040026	.9249331	3.65	0.000	1.674549	5.518954
age		.969571	.0259457	-1.15	0.248	.9200289	1.021781
Asian		(omitted)					
Black		4.491674	3.847904	1.75	0.080	.8379253	24.07749
Latino		.7326414	.6456537	-0.35	0.724	.130243	4.121245
Other		326902	2.26e+08	0.02	0.985	0	.
AsianMale		(omitted)					
BlackMale		1.51e-07	.0001381	-0.02	0.986	0	.
LatinoMale		2.67974	2.87126	0.92	0.358	.3281377	21.88413
OtherMale		(omitted)					

-----> region = 4

Iteration 0: log likelihood = -4431.908  
Iteration 1: log likelihood = -4184.8621  
Iteration 2: log likelihood = -4175.0568  
Iteration 3: log likelihood = -4175.0319  
Iteration 4: log likelihood = -4175.0319

Multinomial logistic regression	Number of obs	=	4304
	LR chi2(36)	=	513.75
	Prob > chi2	=	0.0000
Log likelihood = -4175.0319	Pseudo R2	=	0.0580

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
tenure		1.00907	.0060836	1.50	0.134	.9972171 1.021065
exper		1.022937	.0086	2.70	0.007	1.006219 1.039932
pyperstudlep		1.00273	.0045373	0.60	0.547	.9938763 1.011662
pyperstude~s		.9974713	.0026329	-0.96	0.337	.9923241 1.002645
pyperstudw~e		.9912704	.0067644	-1.28	0.199	.9781006 1.004617
pyperstudb~k		.9858729	.0073243	-1.92	0.055	.9716215 1.000333
pyperstudl~o		.9913619	.0070059	-1.23	0.220	.9777252 1.005189
pyperstudm~e		.9998825	.0031334	-0.04	0.970	.9937599 1.006043
male		1.194531	.1124347	1.89	0.059	.9932951 1.436535
age		.9564472	.0067629	-6.30	0.000	.9432836 .9697946
Asian		.7978304	.5598951	-0.32	0.748	.2016327 3.156895
Black		.966217	.115058	-0.29	0.773	.7650906 1.220215
Latino		1.270821	.2194946	1.39	0.165	.9058703 1.782801
Other		.7012428	.4200103	-0.59	0.553	.2167905 2.26828
AsianMale		.3988409	.5180186	-0.71	0.479	.0312788 5.085679
BlackMale		1.655006	.2945872	2.83	0.005	1.167583 2.345911
LatinoMale		1.246369	.338556	0.81	0.417	.7318658 2.122569
OtherMale		1.717282	1.90255	0.49	0.625	.1957981 15.06173
3						
tenure		1.033669	.0068312	5.01	0.000	1.020366 1.047145
exper		1.080441	.0113849	7.34	0.000	1.058356 1.102987
pyperstudlep		1.009892	.0053267	1.87	0.062	.999506 1.020387
pyperstude~s		1.026669	.003448	7.84	0.000	1.019933 1.033449
pyperstudw~e		1.001513	.0087185	0.17	0.862	.9845698 1.018747
pyperstudb~k		.9798887	.0095682	-2.08	0.037	.9613137 .9988227
pyperstudl~o		.9720972	.0089599	-3.07	0.002	.9546938 .9898177
pyperstudm~e		1.001349	.0039884	0.34	0.735	.9935626 1.009197
male		1.187195	.1285202	1.59	0.113	.9602294 1.467808
age		.9669023	.0087578	-3.72	0.000	.9498888 .9842205
Asian		.3843239	.4163992	-0.88	0.377	.045968 3.213211

Black		.5240621	.0802603	-4.22	0.000	.3881691	.7075294
Latino		1.722021	.3201763	2.92	0.003	1.196118	2.47915
Other		.7051108	.5578646	-0.44	0.659	.149558	3.324337
AsianMale		1.050545	1.624944	0.03	0.975	.0506766	21.7782
BlackMale		1.693225	.3860554	2.31	0.021	1.08303	2.647211
LatinoMale		1.023745	.319942	0.08	0.940	.5548467	1.888906
OtherMale		3.716794	4.577967	1.07	0.286	.3324646	41.55197

-----> region = 5

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -554.08847  
Iteration 1: log likelihood = -495.8567  
Iteration 2: log likelihood = -493.7632  
Iteration 3: log likelihood = -493.59838  
Iteration 4: log likelihood = -493.57293  
Iteration 5: log likelihood = -493.56676  
Iteration 6: log likelihood = -493.56548  
Iteration 7: log likelihood = -493.5652  
Iteration 8: log likelihood = -493.56514  
Iteration 9: log likelihood = -493.56513

Multinomial logistic regression	Number of obs	=	552
	LR chi2(32)	=	121.05
	Prob > chi2	=	0.0000
Log likelihood = -493.56513	Pseudo R2	=	0.1092

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.021915	.0182611	1.21	0.225	.9867437 1.058341
exper		.9918577	.0266674	-0.30	0.761	.9409437 1.045527
pyperstudlep		.9596842	.041111	-0.96	0.337	.882398 1.04374
pyperstude~s		1.012775	.0090038	1.43	0.153	.9952808 1.030577
pyperstudw~e		.9863178	.0384813	-0.35	0.724	.9137073 1.064698
pyperstudb~k		.9730454	.0387842	-0.69	0.493	.8999231 1.052109
pyperstudl~o		1.000937	.0437383	0.02	0.983	.91878 1.090441
pyperstudm~e		.9984191	.0187361	-0.08	0.933	.9623641 1.035825
male		3.085023	.8735212	3.98	0.000	1.771094 5.373723
age		.9671587	.0221492	-1.46	0.145	.924707 1.011559
Asian		(omitted)				
Black		2.233716	.9034897	1.99	0.047	1.010963 4.935379
Latino		3.718203	4.754199	1.03	0.304	.3033631 45.57256
Other		1.12e-06	.0020117	-0.01	0.994	0 .
AsianMale		(omitted)				
BlackMale		.9586758	.5033147	-0.08	0.936	.3425959 2.682634
LatinoMale		2.48e-07	.0002231	-0.02	0.987	0 .
OtherMale		9.85e+12	3.13e+16	0.01	0.992	0 .
3						
tenure		1.031869	.0167674	1.93	0.054	.9995234 1.065261
exper		1.052342	.0301638	1.78	0.075	.9948522 1.113154
pyperstudlep		.9847584	.0414616	-0.36	0.715	.9067577 1.069469
pyperstude~s		1.046988	.0095404	5.04	0.000	1.028455 1.065855
pyperstudw~e		1.018171	.0383966	0.48	0.633	.9456293 1.096278
pyperstudb~k		.9776213	.037844	-0.58	0.559	.9061924 1.05468
pyperstudl~o		.9938044	.0436228	-0.14	0.887	.9118798 1.083089
pyperstudm~e		1.010653	.0175037	0.61	0.541	.9769225 1.045549
male		4.763957	1.36887	5.43	0.000	2.712592 8.366643
age		.9626609	.0248156	-1.48	0.140	.9152315 1.012548
Asian		(omitted)				

Black		2.103961	.9388933	1.67	0.096	.8773818	5.045296
Latino		8.56e-06	.0073364	-0.01	0.989	0	.
Other		2.76e-06	.0047393	-0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		.7057475	.4106435	-0.60	0.549	.2256174	2.20763
LatinoMale		55247.93	4.74e+07	0.01	0.990	0	.
OtherMale		528854.1	2.68e+09	0.00	0.998	0	.

-----> region = 6

```

Iteration 0: log likelihood = -707.63253
Iteration 1: log likelihood = -667.97373
Iteration 2: log likelihood = -667.18773
Iteration 3: log likelihood = -667.08962
Iteration 4: log likelihood = -667.07085
Iteration 5: log likelihood = -667.06681
Iteration 6: log likelihood = -667.06588
Iteration 7: log likelihood = -667.06566
Iteration 8: log likelihood = -667.06561
Iteration 9: log likelihood = -667.0656

```

Multinomial logistic regression	Number of obs	=	680
	LR chi2(36)	=	81.13
	Prob > chi2	=	0.0000
Log likelihood = -667.0656	Pseudo R2	=	0.0573

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9547062	.0148872	-2.97	0.003	.9259692 .984335
exper		.9988459	.0201086	-0.06	0.954	.9602012 1.039046
pyperstudlep		1.033894	.0229284	1.50	0.133	.9899182 1.079824
pyperstude~s		1.006584	.0082687	0.80	0.424	.9905077 1.022922
pyperstudw~e		.971924	.065961	-0.42	0.675	.8508721 1.110198
pyperstudb~k		.9603793	.0671044	-0.58	0.563	.8374655 1.101333
pyperstudl~o		.9711997	.0657846	-0.43	0.666	.8504565 1.109085
pyperstudm~e		.9959413	.0106795	-0.38	0.704	.9752282 1.017094
male		1.495355	.3343667	1.80	0.072	.9647454 2.3178
age		.9960123	.0180908	-0.22	0.826	.9611787 1.032108
Asian		5.07e-07	.0015102	-0.00	0.996	0 .
Black		1.403563	.8123273	0.59	0.558	.4514294 4.363892
Latino		.1300738	.1437039	-1.85	0.065	.0149208 1.133931
Other		5.45e-07	.0016214	-0.00	0.996	0 .
AsianMale		1881760	1.28e+10	0.00	0.998	0 .
BlackMale		.9399043	.7572729	-0.08	0.939	.1937661 4.559209
LatinoMale		11.05575	15.35734	1.73	0.084	.7264389 168.2585
OtherMale		.8306509	2862.725	-0.00	1.000	0 .
3						
tenure		.9807819	.0131001	-1.45	0.146	.9554393 1.006797
exper		1.05296	.0219759	2.47	0.013	1.010757 1.096925
pyperstudlep		1.010456	.0212383	0.49	0.621	.9696754 1.052952
pyperstude~s		1.032139	.0077553	4.21	0.000	1.01705 1.047452
pyperstudw~e		.9191234	.0527185	-1.47	0.141	.8213933 1.028481
pyperstudb~k		.9093596	.0533443	-1.62	0.105	.8105933 1.02016
pyperstudl~o		.9082654	.0519689	-1.68	0.093	.8119119 1.016054
pyperstudm~e		.999094	.0095121	-0.10	0.924	.9806234 1.017912
male		1.813671	.3799474	2.84	0.004	1.202928 2.734496
age		.9800477	.0187874	-1.05	0.293	.9439082 1.017571
Asian		4.13e-07	.0010988	-0.01	0.996	0 .
Black		.7249597	.4493172	-0.52	0.604	.2151564 2.442719
Latino		.1746965	.1915099	-1.59	0.111	.0203779 1.497642

Other		5.10e-07	.0013583	-0.01	0.996	0	.
AsianMale		8.38e+12	3.63e+16	0.01	0.995	0	.
BlackMale		1.286941	1.051841	0.31	0.758	.2593317	6.38648
LatinoMale		7.875029	10.88273	1.49	0.135	.5247501	118.1821
OtherMale		3045011	8.11e+09	0.01	0.996	0	.

-----> region = 7

```

Iteration 0: log likelihood = -1011.5112
Iteration 1: log likelihood = -915.34037
Iteration 2: log likelihood = -908.31718
Iteration 3: log likelihood = -907.82038
Iteration 4: log likelihood = -907.77387
Iteration 5: log likelihood = -907.76426
Iteration 6: log likelihood = -907.76222
Iteration 7: log likelihood = -907.76176
Iteration 8: log likelihood = -907.76165
Iteration 9: log likelihood = -907.76163

```

Multinomial logistic regression	Number of obs	=	1061
	LR chi2(36)	=	207.50
	Prob > chi2	=	0.0000
Log likelihood = -907.76163	Pseudo R2	=	0.1026

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9782858	.0143072	-1.50	0.133	.9506422 1.006733
exper		1.008943	.0188593	0.48	0.634	.9726483 1.046592
pyperstudlep		.9824275	.0214171	-0.81	0.416	.941335 1.025314
pyperstude~s		.9931603	.0083067	-0.82	0.412	.9770123 1.009575
pyperstudw~e		.7982969	.1176138	-1.53	0.126	.5980757 1.065547
pyperstudb~k		.8028579	.1186218	-1.49	0.137	.6009995 1.072515
pyperstudl~o		.8297929	.1229211	-1.26	0.208	.6206935 1.109334
pyperstudm~e		.9251098	.0233168	-3.09	0.002	.8805202 .9719575
male		2.319512	.5335454	3.66	0.000	1.477743 3.640777
age		.9800634	.0163287	-1.21	0.227	.9485767 1.012595
Asian		7.84e+08	4.60e+12	0.00	0.997	0 .
Black		2.148204	.813747	2.02	0.044	1.022439 4.513502
Latino		.8467654	.9683931	-0.15	0.884	.0900112 7.965803
Other		3.039587	3.798596	0.89	0.374	.2624589 35.20205
AsianMale		1.28e-09	8.41e-06	-0.00	0.998	0 .
BlackMale		1.74573	.9266531	1.05	0.294	.6168078 4.940881
LatinoMale		2.773747	4.18457	0.68	0.499	.1441834 53.36033
OtherMale		886235.7	6.07e+08	0.02	0.984	0 .
3						
tenure		.9751516	.0101187	-2.42	0.015	.9555197 .9951869
exper		1.085924	.0167706	5.34	0.000	1.053547 1.119297
pyperstudlep		1.021207	.0184108	1.16	0.244	.9857531 1.057937
pyperstude~s		1.043266	.00632	6.99	0.000	1.030952 1.055727
pyperstudw~e		.9071406	.1157149	-0.76	0.445	.7064726 1.164807
pyperstudb~k		.8865619	.1136013	-0.94	0.347	.6896658 1.139671
pyperstudl~o		.8769251	.1139345	-1.01	0.312	.6797827 1.13124
pyperstudm~e		1.008041	.009084	0.89	0.374	.9903925 1.026003
male		1.989895	.3337328	4.10	0.000	1.432425 2.76432
age		.9649334	.0136342	-2.53	0.012	.9385775 .9920294
Asian		2.650579	26680.07	0.00	1.000	0 .
Black		.7322191	.2522513	-0.90	0.366	.3727348 1.438408
Latino		.1972388	.2203127	-1.45	0.146	.0220909 1.761049
Other		2.81e-06	.0018974	-0.02	0.985	0 .
AsianMale		545010.5	5.53e+09	0.00	0.999	0 .

BlackMale		3.118612	1.548623	2.29	0.022	1.178356	8.253655
LatinoMale		6.102967	9.188799	1.20	0.230	.3191156	116.717
OtherMale		7.33e+11	7.05e+14	0.03	0.977	0	.

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -313.08709  
Iteration 1: log likelihood = -272.25817  
Iteration 2: log likelihood = -268.55963  
Iteration 3: log likelihood = -268.31144  
Iteration 4: log likelihood = -268.26942  
Iteration 5: log likelihood = -268.25985  
Iteration 6: log likelihood = -268.25757  
Iteration 7: log likelihood = -268.25711  
Iteration 8: log likelihood = -268.25704  
Iteration 9: log likelihood = -268.25702

Multinomial logistic regression	Number of obs	=	316
	LR chi2(34)	=	89.66
	Prob > chi2	=	0.0000
Log likelihood = -268.25702	Pseudo R2	=	0.1432

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.00127	.0270866	0.05	0.963	.9495643 1.055791
exper		1.074398	.0534255	1.44	0.149	.9746264 1.184382
pyperstudlep		.8998915	.0688795	-1.38	0.168	.7745287 1.045545
pyperstude~s		.9804459	.0177165	-1.09	0.274	.94633 1.015792
pyperstudw~e		.7867016	.2140176	-0.88	0.378	.4615785 1.340832
pyperstudb~k		.8216583	.2226287	-0.72	0.468	.483123 1.397413
pyperstudl~o		.8493494	.2312982	-0.60	0.549	.4980624 1.448402
pyperstudm~e		.9349074	.0403519	-1.56	0.119	.8590719 1.017437
male		3.100178	1.389151	2.53	0.012	1.288172 7.46104
age		.9002509	.0416097	-2.27	0.023	.8222823 .9856125
Asian		2.54e+08	1.83e+12	0.00	0.998	0 .
Black		21.81572	21.36089	3.15	0.002	3.201215 148.6704
Latino		4.616637	6.445117	1.10	0.273	.2992228 71.229
Other		7.92e-07	.0008871	-0.01	0.990	0 .
AsianMale		(omitted)				
BlackMale		.1764468	.2193191	-1.40	0.163	.015438 2.016679
LatinoMale		1.34e+07	9.68e+10	0.00	0.998	0 .
OtherMale		1985762	2.22e+09	0.01	0.990	0 .
3						
tenure		.976687	.0175134	-1.32	0.188	.9429575 1.011623
exper		1.095481	.0345759	2.89	0.004	1.029767 1.165388
pyperstudlep		.9534509	.0446105	-1.02	0.308	.8699052 1.04502
pyperstude~s		1.037843	.0110255	3.50	0.000	1.016456 1.059679
pyperstudw~e		1.012034	.1860866	0.07	0.948	.7057993 1.451138
pyperstudb~k		.9980009	.1837503	-0.01	0.991	.6956793 1.431702
pyperstudl~o		1.03202	.1907574	0.17	0.865	.7183781 1.482598
pyperstudm~e		1.006718	.0162959	0.41	0.679	.9752801 1.03917
male		2.45123	.7084474	3.10	0.002	1.391145 4.319126
age		.965812	.0273537	-1.23	0.219	.9136605 1.02094
Asian		1.3009	14610.39	0.00	1.000	0 .
Black		4.897858	4.483567	1.74	0.083	.8143377 29.45831
Latino		.6892285	.8340728	-0.31	0.758	.0643099 7.386666
Other		7.31e-07	.0004807	-0.02	0.983	0 .
AsianMale		(omitted)				
BlackMale		.3214118	.3834606	-0.95	0.341	.0310124 3.331107



LatinoMale		1.128423	12673.31	0.00	1.000	0	.
OtherMale		1425329	9.37e+08	0.02	0.983	0	.

---

-----> region = 9

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -207.39036  
Iteration 1: log likelihood = -186.98811  
Iteration 2: log likelihood = -185.36269  
Iteration 3: log likelihood = -184.99509  
Iteration 4: log likelihood = -184.91041  
Iteration 5: log likelihood = -184.89309  
Iteration 6: log likelihood = -184.8894  
Iteration 7: log likelihood = -184.88857  
Iteration 8: log likelihood = -184.88836  
Iteration 9: log likelihood = -184.88832  
Iteration 10: log likelihood = -184.88831

Multinomial logistic regression		Number of obs	=	210
		LR chi2(28)	=	45.00
		Prob > chi2	=	0.0220
Log likelihood = -184.88831		Pseudo R2	=	0.1085

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.014359	.0356615	0.41	0.685	.9468174 1.086719
exper		.994171	.0529868	-0.11	0.913	.895559 1.103641
pyperstudlep		.9877423	.1171549	-0.10	0.917	.7828592 1.246246
pyperstude~s		.9727837	.020428	-1.31	0.189	.9335583 1.013657
pyperstudw~e		.9535646	.198436	-0.23	0.819	.6341847 1.433786
pyperstudb~k		1.018804	.2313527	0.08	0.935	.652826 1.589951
pyperstudl~o		.9559802	.1993638	-0.22	0.829	.6352375 1.438672
pyperstudm~e		.9644676	.0395497	-0.88	0.378	.8899849 1.045184
male		1.659143	.8304728	1.01	0.312	.6220435 4.425344
age		.97119	.0431375	-0.66	0.510	.8902177 1.059527
Asian		(omitted)				
Black		3816159	3.51e+09	0.02	0.987	0 .
Latino		13.71486	59376.92	0.00	1.000	0 .
Other		1.96403	2.940729	0.45	0.652	.1043878 36.95273
AsianMale		(omitted)				
BlackMale		2.27e-14	3.51e-11	-0.02	0.984	0 .
LatinoMale		(omitted)				
OtherMale		(omitted)				
3						
tenure		1.00442	.0218058	0.20	0.839	.9625774 1.04808
exper		1.043712	.0369028	1.21	0.226	.9738334 1.118605
pyperstudlep		.9920128	.0602613	-0.13	0.895	.8806631 1.117441
pyperstude~s		1.028089	.012854	2.22	0.027	1.003202 1.053593
pyperstudw~e		1.350186	.1825486	2.22	0.026	1.035879 1.75986
pyperstudb~k		1.368586	.1965314	2.19	0.029	1.032852 1.813452
pyperstudl~o		1.318599	.1774082	2.06	0.040	1.012954 1.716469
pyperstudm~e		.9891722	.0218103	-0.49	0.621	.9473352 1.032857
male		1.931052	.6512408	1.95	0.051	.9970776 3.73989
age		.9884848	.0308984	-0.37	0.711	.9297428 1.050938
Asian		(omitted)				
Black		2323530	2.14e+09	0.02	0.987	0 .
Latino		8434328	1.73e+10	0.01	0.994	0 .

Other		2.74e-07	.0006799	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		1.87e-07	.0001717	-0.02	0.987	0	.
LatinoMale		(omitted)					
OtherMale		(omitted)					

-----> region = 10

Iteration 0: log likelihood = -2848.908  
Iteration 1: log likelihood = -2703.204  
Iteration 2: log likelihood = -2701.0001  
Iteration 3: log likelihood = -2700.9795  
Iteration 4: log likelihood = -2700.9795

Multinomial logistic regression	Number of obs	=	2699
	LR chi2(36)	=	295.86
	Prob > chi2	=	0.0000
Log likelihood = -2700.9795	Pseudo R2	=	0.0519

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.011445	.0075605	1.52	0.128	.9967348 1.026373
exper		1.011582	.0102715	1.13	0.257	.9916491 1.031915
pyperstudlep		.9973563	.006523	-0.40	0.686	.984653 1.010224
pyperstude~s		.9956826	.0037534	-1.15	0.251	.9883531 1.003066
pyperstudw~e		.9892725	.011142	-0.96	0.338	.9676738 1.011353
pyperstudb~k		.9909325	.0114711	-0.79	0.431	.9687026 1.013672
pyperstudl~o		.9947459	.0114264	-0.46	0.647	.9726008 1.017395
pyperstudm~e		.9955134	.005048	-0.89	0.375	.9856685 1.005457
male		1.559895	.1855439	3.74	0.000	1.235515 1.969439
age		.9661418	.0084316	-3.95	0.000	.9497567 .9828096
Asian		.5330736	.5898532	-0.57	0.570	.0609426 4.662869
Black		2.239825	.3931782	4.59	0.000	1.587795 3.159612
Latino		3.287492	.8435997	4.64	0.000	1.988102 5.436139
Other		1.205942	.7254576	0.31	0.756	.3709113 3.920875
AsianMale		2.233874	2.788187	0.64	0.520	.1934819 25.79153
BlackMale		1.125276	.2976598	0.45	0.655	.6700353 1.889819
LatinoMale		1.228428	.4569581	0.55	0.580	.5925356 2.546741
OtherMale		.5878194	.5322101	-0.59	0.557	.0996709 3.466726
3						
tenure		1.016657	.0070088	2.40	0.017	1.003013 1.030487
exper		1.067593	.0116247	6.01	0.000	1.04505 1.090622
pyperstudlep		1.00796	.0069711	1.15	0.252	.994389 1.021716
pyperstude~s		1.016147	.0037711	4.32	0.000	1.008783 1.023566
pyperstudw~e		1.023176	.0119597	1.96	0.050	1.000002 1.046887
pyperstudb~k		1.004662	.0122723	0.38	0.703	.9808941 1.029005
pyperstudl~o		.9964264	.0121182	-0.29	0.768	.972956 1.020463
pyperstudm~e		.9956657	.0052538	-0.82	0.410	.9854216 1.006016
male		1.254247	.1453241	1.96	0.051	.999443 1.574012
age		.9701736	.0093044	-3.16	0.002	.9521076 .9885824
Asian		.6047523	.5338843	-0.57	0.569	.1071826 3.412172
Black		1.399741	.2626962	1.79	0.073	.9689441 2.022073
Latino		2.511832	.6660666	3.47	0.001	1.493744 4.223817
Other		.2342353	.2501045	-1.36	0.174	.0288923 1.898987
AsianMale		2.412441	2.477019	0.86	0.391	.3224556 18.0486
BlackMale		1.388416	.4012092	1.14	0.256	.7880406 2.446191
LatinoMale		1.076351	.4455192	0.18	0.859	.4782199 2.422591
OtherMale		3.29528	4.194583	0.94	0.349	.2718895 39.93855

-----> region = 11

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -2029.57  
 Iteration 1: log likelihood = -1869.6665  
 Iteration 2: log likelihood = -1867.4381  
 Iteration 3: log likelihood = -1867.3024  
 Iteration 4: log likelihood = -1867.2915  
 Iteration 5: log likelihood = -1867.2893  
 Iteration 6: log likelihood = -1867.289  
 Iteration 7: log likelihood = -1867.2889  
 Iteration 8: log likelihood = -1867.2889

Multinomial logistic regression

Number of obs = 1887  
 LR chi2(34) = 324.56  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0800

Log likelihood = -1867.2889

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.964337	.0093243	-3.76	0.000	.9462338	.9827865
	exper	.9419939	.0093161	-6.04	0.000	.9239106	.9604312
pyperstudlep		1.007302	.0115425	0.63	0.525	.9849309	1.030181
pyperstude~s		.9967623	.004472	-0.72	0.470	.9880358	1.005566
pyperstudw~e		.9572891	.0166379	-2.51	0.012	.9252287	.9904605
pyperstudb~k		.9510396	.0172665	-2.76	0.006	.9177929	.9854906
pyperstudl~o		.9456253	.0168292	-3.14	0.002	.9132093	.9791919
pyperstudm~e		.9959791	.0048381	-0.83	0.407	.9865417	1.005507
	male	.8422004	.11184	-1.29	0.196	.6492025	1.092574
	age	1.030945	.0087556	3.59	0.000	1.013926	1.048249
	Asian	1.642972	1.449308	0.56	0.574	.2915837	9.25757
	Black	1.167743	.3105372	0.58	0.560	.6934058	1.966558
	Latino	2.032406	.7468121	1.93	0.054	.9890873	4.176249
	Other	1.268232	.9866709	0.31	0.760	.2760406	5.826723
	AsianMale	(omitted)					
	BlackMale	3.409898	1.417065	2.95	0.003	1.510095	7.699785
	LatinoMale	.7517195	.3987215	-0.54	0.591	.2658078	2.125905
	OtherMale	.3722902	.532966	-0.69	0.490	.0225067	6.158163
3							
	tenure	.9827587	.0083537	-2.05	0.041	.9665214	.9992689
	exper	.9688284	.0094216	-3.26	0.001	.9505372	.9874716
pyperstudlep		1.010825	.0120944	0.90	0.368	.9873962	1.03481
pyperstude~s		1.028541	.0044434	6.51	0.000	1.019868	1.037286
pyperstudw~e		1.029745	.0191272	1.58	0.115	.9929308	1.067925
pyperstudb~k		.9991093	.0194225	-0.05	0.963	.9617579	1.037911
pyperstudl~o		.9848491	.0192679	-0.78	0.435	.9477996	1.023347
pyperstudm~e		1.003956	.0045899	0.86	0.388	.995	1.012992
	male	1.306888	.1663578	2.10	0.035	1.018325	1.677222
	age	1.063191	.0095654	6.81	0.000	1.044608	1.082105
	Asian	3.52e-06	.0018013	-0.02	0.980	0	.
	Black	1.586716	.453376	1.62	0.106	.906324	2.77789
	Latino	3.685291	1.373514	3.50	0.000	1.775119	7.650961
	Other	1.137449	.9930027	0.15	0.883	.2055066	6.295615
	AsianMale	(omitted)					
	BlackMale	1.134017	.5377268	0.27	0.791	.4477109	2.872376
	LatinoMale	.2638492	.158804	-2.21	0.027	.0811035	.8583648
	OtherMale	.2438238	.3580168	-0.96	0.336	.0137158	4.334425

-----> region = 12

Multinomial logistic regression	Number of obs	=	728
	LR chi2(32)	=	129.37
	Prob > chi2	=	0.0000
Log likelihood = -700.16749	Pseudo R2	=	0.0846

```
-----> region = 13
```

```

Iteration 0: log likelihood = -1189.4141
Iteration 1: log likelihood = -1128.004
Iteration 2: log likelihood = -1126.8106
Iteration 3: log likelihood = -1126.6107
Iteration 4: log likelihood = -1126.5769
Iteration 5: log likelihood = -1126.5692
Iteration 6: log likelihood = -1126.5674
Iteration 7: log likelihood = -1126.567
Iteration 8: log likelihood = -1126.5669
Iteration 9: log likelihood = -1126.5669

```

```

Multinomial logistic regression      Number of obs   =      1100
                                     LR chi2(36)        =      125.69
                                     Prob > chi2         =      0.0000
Log likelihood = -1126.5669          Pseudo R2        =      0.0528

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013186	.0117953	1.13	0.261	.990329	1.03657
	exper	1.004179	.0155896	0.27	0.788	.9740843	1.035204
pyperstudlep		.9878146	.0142145	-0.85	0.394	.9603439	1.016071
pyperstude~s		.993329	.0074361	-0.89	0.371	.978861	1.008011
pyperstudw~e		.934562	.0302139	-2.09	0.036	.8771809	.9956966
pyperstudb~k		.9561532	.0329611	-1.30	0.193	.8936847	1.022988
pyperstudl~o		.9484858	.030251	-1.66	0.097	.89101	1.009669
pyperstudm~e		.9946328	.0066478	-0.81	0.421	.9816883	1.007748
	male	1.036444	.1834118	0.20	0.840	.7326812	1.466144
	age	.9672393	.0137957	-2.34	0.020	.9405746	.99466
	Asian	1.836397	5113.292	0.00	1.000	0	.
	Black	2.23068	.8505689	2.10	0.035	1.056502	4.709822
	Latino	1.251977	.3652762	0.77	0.441	.7067245	2.217903
	Other	6.72e-07	.000554	-0.02	0.986	0	.
	AsianMale	1809981	5.99e+09	0.00	0.997	0	.
	BlackMale	.9536057	.5692567	-0.08	0.937	.2959665	3.072523
	LatinoMale	2.483012	1.142299	1.98	0.048	1.007829	6.117451
	OtherMale	4199374	3.46e+09	0.02	0.985	0	.
3							
	tenure	.9908536	.010988	-0.83	0.407	.96955	1.012625
	exper	1.055356	.0165075	3.44	0.001	1.023493	1.088211
pyperstudlep		.9840997	.0137827	-1.14	0.252	.9574535	1.011488
pyperstude~s		1.033477	.0073597	4.62	0.000	1.019152	1.048003
pyperstudw~e		.9746461	.0326681	-0.77	0.444	.9126755	1.040824
pyperstudb~k		.9623287	.0344927	-1.07	0.284	.8970442	1.032364
pyperstudl~o		.9554075	.0318598	-1.37	0.171	.8949604	1.019937
pyperstudm~e		.9946691	.0068519	-0.78	0.438	.9813299	1.00819
	male	1.261627	.2113351	1.39	0.165	.9085449	1.751926
	age	.9773234	.0143432	-1.56	0.118	.9496116	1.005844
	Asian	7222629	1.28e+10	0.01	0.993	0	.
	Black	1.269976	.5624303	0.54	0.589	.533122	3.025272
	Latino	1.15671	.347758	0.48	0.628	.6416757	2.08513
	Other	3.98e-07	.0003625	-0.02	0.987	0	.
	AsianMale	2.39e-07	.0007929	-0.00	0.996	0	.
	BlackMale	1.14205	.7701844	0.20	0.844	.3045404	4.282772
	LatinoMale	1.024505	.5322493	0.05	0.963	.3700828	2.836149
	OtherMale	8.312232	10148.55	0.00	0.999	0	.

-----> region = 14

note: Asian omitted because of collinearity

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -284.31818  
 Iteration 1: log likelihood = -250.31581  
 Iteration 2: log likelihood = -246.5636  
 Iteration 3: log likelihood = -246.27646  
 Iteration 4: log likelihood = -246.23298  
 Iteration 5: log likelihood = -246.22265  
 Iteration 6: log likelihood = -246.22028  
 Iteration 7: log likelihood = -246.21978  
 Iteration 8: log likelihood = -246.21969  
 Iteration 9: log likelihood = -246.21968

Multinomial logistic regression	Number of obs	=	301
	LR chi2(32)	=	76.20
	Prob > chi2	=	0.0000
Log likelihood = -246.21968	Pseudo R2	=	0.1340

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9913852	.0331992	-0.26	0.796	.9284053	1.058637
exper	1.027169	.0569561	0.48	0.629	.9213888	1.145092
pyperstudlep	.8303962	.1245315	-1.24	0.215	.618919	1.114133
pyperstude~s	.999554	.0209525	-0.02	0.983	.95932	1.041475
pyperstudw~e	.721135	.1786087	-1.32	0.187	.4438062	1.171763
pyperstudb~k	.7769295	.2043686	-0.96	0.337	.4639551	1.30103
pyperstudl~o	.7388094	.1764688	-1.27	0.205	.4626132	1.179904
pyperstudm~e	.9709314	.0323588	-0.89	0.376	.9095363	1.036471
male	2.258592	1.087355	1.69	0.091	.8791101	5.802727
age	.9449023	.0462991	-1.16	0.247	.8583789	1.040147
Asian	(omitted)					
Black	9.16e-07	.001733	-0.01	0.994	0	.
Latino	8.53678	16.13798	1.13	0.257	.2099798	347.0649
Other	6.09e-06	.0136941	-0.01	0.996	0	.
AsianMale	(omitted)					
BlackMale	.405781	1072.616	-0.00	1.000	0	.
LatinoMale	.2955994	.6443993	-0.56	0.576	.004122	21.19833
OtherMale	.2071425	629.8468	-0.00	1.000	0	.
3						
tenure	.9743611	.0195622	-1.29	0.196	.9367644	1.013467
exper	1.088021	.0348956	2.63	0.009	1.021732	1.15861
pyperstudlep	1.115996	.0646659	1.89	0.058	.9961855	1.250217
pyperstude~s	1.037605	.0116018	3.30	0.001	1.015113	1.060595
pyperstudw~e	1.16249	.1990683	0.88	0.379	.8310497	1.626116
pyperstudb~k	1.109387	.2063724	0.56	0.577	.7704407	1.597448
pyperstudl~o	1.124319	.189487	0.70	0.487	.8080392	1.564396
pyperstudm~e	1.005583	.0179979	0.31	0.756	.9709193	1.041484
male	2.344782	.6801394	2.94	0.003	1.328002	4.140057
age	.9346889	.0276424	-2.28	0.022	.8820511	.990468
Asian	(omitted)					
Black	1.557854	2.189151	0.32	0.752	.0991719	24.47174
Latino	2.15e-07	.0002723	-0.01	0.990	0	.
Other	1.573485	2.345915	0.30	0.761	.0846842	29.2363
AsianMale	(omitted)					
BlackMale	1.120747	2.105265	0.06	0.952	.0282208	44.50873
LatinoMale	3992504	5.06e+09	0.01	0.990	0	.
OtherMale	.2736107	.5737759	-0.62	0.537	.0044887	16.67812

-----> region = 15

note: Black omitted because of collinearity

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -304.55879  
Iteration 1: log likelihood = -266.38427  
Iteration 2: log likelihood = -264.18073  
Iteration 3: log likelihood = -264.12439  
Iteration 4: log likelihood = -264.11616  
Iteration 5: log likelihood = -264.11472  
Iteration 6: log likelihood = -264.11457  
Iteration 7: log likelihood = -264.11454  
Iteration 8: log likelihood = -264.11453

Multinomial logistic regression	Number of obs	=	317
	LR chi2(26)	=	80.89
	Prob > chi2	=	0.0000
Log likelihood = -264.11453	Pseudo R2	=	0.1328

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.003922	.0290879	0.14	0.893	.9484996	1.062583
	exper	1.161515	.0698798	2.49	0.013	1.03232	1.306879
pyperstudlep		.9905586	.0268216	-0.35	0.726	.9393597	1.044548
pyperstude~s		.9960324	.0159036	-0.25	0.803	.9653447	1.027696
pyperstudw~e		1.171625	.4214659	0.44	0.660	.578879	2.371315
pyperstudb~k		1.315526	.523183	0.69	0.490	.6033614	2.868278
pyperstudl~o		1.20495	.4372258	0.51	0.607	.5917029	2.453773
pyperstudm~e		1.004458	.0212686	0.21	0.834	.9636259	1.047021
	male	1.904801	.8884565	1.38	0.167	.7635259	4.751991
	age	.842325	.0469146	-3.08	0.002	.7552152	.9394824
	Asian	1.36e-06	.0020879	-0.01	0.993	0	.
	Black	(omitted)					
	Latino	.3619193	.3253875	-1.13	0.258	.0621341	2.108112
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	2.94683	2.951733	1.08	0.281	.4137495	20.98808
	OtherMale	(omitted)					
3							
	tenure	1.002062	.0195445	0.11	0.916	.9644785	1.04111
	exper	1.065999	.0328658	2.07	0.038	1.003491	1.132401
pyperstudlep		1.014404	.0217562	0.67	0.505	.972647	1.057955
pyperstude~s		1.040616	.0107682	3.85	0.000	1.019724	1.061937
pyperstudw~e		.9745308	.2153882	-0.12	0.907	.6319246	1.502885
pyperstudb~k		.9020477	.2145652	-0.43	0.665	.5659249	1.437806
pyperstudl~o		.9361023	.2071754	-0.30	0.765	.6066496	1.444471
pyperstudm~e		1.017988	.0164248	1.10	0.269	.9862992	1.050694
	male	2.557987	.794926	3.02	0.003	1.39116	4.703485
	age	.9794432	.0267174	-0.76	0.446	.9284533	1.033233
	Asian	.0000106	.0094556	-0.01	0.990	0	.
	Black	(omitted)					
	Latino	.9723297	.5988142	-0.05	0.964	.2908016	3.2511
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.568701	1.120752	0.63	0.529	.3867252	6.363232
	OtherMale	(omitted)					

-----> region = 16

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -456.51647  
 Iteration 1: log likelihood = -409.76386  
 Iteration 2: log likelihood = -405.95285  
 Iteration 3: log likelihood = -405.44912  
 Iteration 4: log likelihood = -405.34785  
 Iteration 5: log likelihood = -405.32614  
 Iteration 6: log likelihood = -405.32248  
 Iteration 7: log likelihood = -405.32162  
 Iteration 8: log likelihood = -405.32142  
 Iteration 9: log likelihood = -405.32138  
 Iteration 10: log likelihood = -405.32137

Multinomial logistic regression

Number of obs = 433  
 LR chi2(34) = 102.39  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1121

Log likelihood = -405.32137

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.065205	.0266882	2.52	0.012	1.014161	1.118818
	exper	.9569374	.0300903	-1.40	0.162	.8997421	1.017769
pyperstudlep		.9915916	.0232453	-0.36	0.719	.9470624	1.038214
pyperstude~s		.9856656	.0103111	-1.38	0.168	.9656619	1.006084
pyperstudw~e		.9110006	.0339999	-2.50	0.013	.846741	.9801369
pyperstudb~k		.9391998	.040494	-1.45	0.146	.8630938	1.022017
pyperstudl~o		.9260634	.0331613	-2.15	0.032	.8632969	.9933935
pyperstudm~e		1.002193	.0105451	0.21	0.835	.9817368	1.023076
	male	.9301997	.2855393	-0.24	0.814	.5096671	1.697719
	age	.965588	.0263951	-1.28	0.200	.915216	1.018732
	Asian	1.95e+08	2.00e+12	0.00	0.999	0	.
	Black	.6048325	.6169054	-0.49	0.622	.08193	4.465059
	Latino	3.207425	2.175988	1.72	0.086	.8485603	12.12357
	Other	7.79e-08	.0003227	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	1.219175	1.97546	0.12	0.903	.0509147	29.1937
	LatinoMale	1.271679	1.443242	0.21	0.832	.1375118	11.76021
	OtherMale	2.42e+07	1.00e+11	0.00	0.997	0	.
3							
	tenure	1.032041	.0170111	1.91	0.056	.999233	1.065927
	exper	1.01816	.0239431	0.77	0.444	.9722975	1.066186
pyperstudlep		1.017357	.0190185	0.92	0.357	.9807562	1.055324
pyperstude~s		1.021227	.0080293	2.67	0.008	1.005611	1.037086
pyperstudw~e		1.008374	.042717	0.20	0.844	.9280319	1.095672
pyperstudb~k		.9742139	.0486961	-0.52	0.601	.8832975	1.074488
pyperstudl~o		.9902533	.0410751	-0.24	0.813	.9129333	1.074122
pyperstudm~e		1.001591	.0101761	0.16	0.876	.9818435	1.021736
	male	1.24814	.3064997	0.90	0.367	.7713257	2.019709
	age	1.013823	.0228811	0.61	0.543	.969954	1.059675
	Asian	3.126479	47288.3	0.00	1.000	0	.
	Black	1.55e-07	.0002798	-0.01	0.993	0	.
	Latino	.4478374	.4137675	-0.87	0.385	.0732277	2.738832
	Other	5.82e-08	.0001939	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	.7255227	2041.138	-0.00	1.000	0	.
	LatinoMale	1.30582	2.038634	0.17	0.864	.0612344	27.84654
	OtherMale	1.414049	5665.367	0.00	1.000	0	.

-----> region = 17



```

Iteration 0: log likelihood = -438.2734
Iteration 1: log likelihood = -391.38241
Iteration 2: log likelihood = -388.1882
Iteration 3: log likelihood = -387.85079
Iteration 4: log likelihood = -387.79017
Iteration 5: log likelihood = -387.77686
Iteration 6: log likelihood = -387.77372
Iteration 7: log likelihood = -387.77302
Iteration 8: log likelihood = -387.77287
Iteration 9: log likelihood = -387.77284

```

```

Multinomial logistic regression      Number of obs   =      437
                                     LR chi2(36)        =     101.00
                                     Prob > chi2         =     0.0000
Log likelihood = -387.77284          Pseudo R2        =     0.1152

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9688253	.0209404	-1.47	0.143	.9286401	1.010749
	exper	1.014381	.0340368	0.43	0.670	.949816	1.083334
pyperstudlep		.9661295	.0430724	-0.77	0.440	.8852924	1.054348
pyperstude~s		.9770896	.0135234	-1.67	0.094	.9509405	1.003958
pyperstudw~e		.8975883	.0834991	-1.16	0.245	.7479857	1.077112
pyperstudb~k		.9277928	.0862059	-0.81	0.420	.7733242	1.113116
pyperstudl~o		.9064852	.0825253	-1.08	0.281	.7583476	1.08356
pyperstudm~e		.995967	.0128978	-0.31	0.755	.971006	1.02157
	male	2.371442	.8795281	2.33	0.020	1.146347	4.905789
	age	1.025166	.0330347	0.77	0.441	.9624211	1.092001
	Asian	7.38e-07	.0026622	-0.00	0.997	0	.
	Black	1.28e-06	.001169	-0.01	0.988	0	.
	Latino	1.00286	1.14868	0.00	0.998	.1062357	9.466944
	Other	2.02215	8641.348	0.00	1.000	0	.
	AsianMale	1074987	6.01e+09	0.00	0.998	0	.
	BlackMale	5.21e+12	8.05e+15	0.02	0.985	0	.
	LatinoMale	4.406649	5.865248	1.11	0.265	.3244587	59.84907
	OtherMale	8.86e-07	.0039804	-0.00	0.998	0	.
3							
	tenure	.9870815	.0154038	-0.83	0.405	.9573476	1.017739
	exper	1.044411	.0265619	1.71	0.088	.9936273	1.097791
pyperstudlep		1.032532	.0231991	1.42	0.154	.988049	1.079017
pyperstude~s		1.030117	.010294	2.97	0.003	1.010138	1.050492
pyperstudw~e		1.028924	.1079699	0.27	0.786	.8376501	1.263875
pyperstudb~k		.9997595	.1055274	-0.00	0.998	.8129219	1.229539
pyperstudl~o		1.008837	.1050666	0.08	0.933	.8225674	1.237287
pyperstudm~e		.9859979	.0114236	-1.22	0.224	.9638603	1.008644
	male	2.468169	.6186538	3.60	0.000	1.510145	4.033955
	age	1.007176	.0236702	0.30	0.761	.9618355	1.054654
	Asian	4.95e-07	.0010407	-0.01	0.994	0	.
	Black	5.23977	5.022011	1.73	0.084	.8007293	34.28773
	Latino	1.198021	.7766835	0.28	0.780	.3362219	4.268772
	Other	8142411	1.72e+10	0.01	0.994	0	.
	AsianMale	8.71e+12	2.59e+16	0.01	0.992	0	.
	BlackMale	859857.5	1.07e+09	0.01	0.991	0	.
	LatinoMale	1.836261	1.662552	0.67	0.502	.3113553	10.82961
	OtherMale	3.35e-07	.0007078	-0.01	0.994	0	.

-----> region = 18

note: Asian omitted because of collinearity

Multinomial logistic regression	Number of obs	=	412
	LR chi2(30)	=	80.62
	Prob > chi2	=	0.0000
Log likelihood = -381.56688	Pseudo R2	=	0.0955

```
-----> region = 19
```

```

Iteration 0: log likelihood = -744.23789
Iteration 1: log likelihood = -683.39139
Iteration 2: log likelihood = -680.91929
Iteration 3: log likelihood = -680.80175
Iteration 4: log likelihood = -680.77568
Iteration 5: log likelihood = -680.77031
Iteration 6: log likelihood = -680.76933
Iteration 7: log likelihood = -680.76923
Iteration 8: log likelihood = -680.76921

```

```

Multinomial logistic regression      Number of obs   =      695
                                     LR chi2(36)       =     126.94
                                     Prob > chi2      =     0.0000
Log likelihood = -680.76921         Pseudo R2       =     0.0853

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.022864	.0142593	1.62	0.105	.9952944	1.051197
	exper	1.042172	.0200615	2.15	0.032	1.003585	1.082243
pyperstudlep		1.006607	.0057512	1.15	0.249	.995398	1.017943
pyperstude~s		1.009372	.0080215	1.17	0.240	.9937719	1.025217
pyperstudw~e		.9083585	.1037841	-0.84	0.400	.7261119	1.136347
pyperstudb~k		.8995515	.1131085	-0.84	0.400	.7030674	1.150946
pyperstudl~o		.8977359	.0999745	-0.97	0.333	.7216994	1.116711
pyperstudm~e		.991445	.0084082	-1.01	0.311	.9751014	1.008063
	male	1.152545	.3144841	0.52	0.603	.6751474	1.967512
	age	.9743821	.0157451	-1.61	0.108	.9440059	1.005736
	Asian	2839825	4.27e+09	0.01	0.992	0	.
	Black	1.264409	.895191	0.33	0.740	.3156779	5.064436
	Latino	1.402051	.3665419	1.29	0.196	.8399085	2.34043
	Other	1.35e-06	.0012848	-0.01	0.989	0	.
	AsianMale	4.29e-07	.0010196	-0.01	0.995	0	.
	BlackMale	.6441969	.6634813	-0.43	0.669	.0855728	4.84955
	LatinoMale	1.253679	.4655869	0.61	0.543	.6054388	2.595987
	OtherMale	2257873	8.42e+09	0.00	0.997	0	.
3							
	tenure	1.041352	.0158292	2.67	0.008	1.010784	1.072843
	exper	1.144909	.0301082	5.15	0.000	1.087393	1.205467
pyperstudlep		1.011954	.0065556	1.83	0.067	.9991866	1.024885
pyperstude~s		1.026073	.0101545	2.60	0.009	1.006363	1.04617
pyperstudw~e		.9274487	.1132199	-0.62	0.537	.7300926	1.178153
pyperstudb~k		.8803924	.1183162	-0.95	0.343	.6765237	1.145696
pyperstudl~o		.9026407	.1072401	-0.86	0.389	.7151319	1.139315
pyperstudm~e		1.004624	.0080373	0.58	0.564	.988994	1.020501
	male	.8797797	.2853928	-0.39	0.693	.4658559	1.661484
	age	.9299255	.0209558	-3.22	0.001	.8897468	.9719186
	Asian	1.110755	2796.882	0.00	1.000	0	.
	Black	.3887131	.4424023	-0.83	0.406	.0417695	3.617421
	Latino	1.123054	.3390942	0.38	0.701	.6214261	2.029607
	Other	2.21e-06	.0023353	-0.01	0.990	0	.
	AsianMale	2.79e+07	7.56e+10	0.01	0.995	0	.
	BlackMale	4.084423	5.604532	1.03	0.305	.2774186	60.1348
	LatinoMale	1.648477	.7220359	1.14	0.254	.6986426	3.889652
	OtherMale	1.51e+14	3.93e+17	0.01	0.990	0	.

```

-----> region = 20

```

```

Iteration 0: log likelihood = -2004.6366
Iteration 1: log likelihood = -1884.9031
Iteration 2: log likelihood = -1880.0141

```

```

Iteration 3: log likelihood = -1879.8204
Iteration 4: log likelihood = -1879.7836
Iteration 5: log likelihood = -1879.7757
Iteration 6: log likelihood = -1879.7739
Iteration 7: log likelihood = -1879.7735
Iteration 8: log likelihood = -1879.7734
Iteration 9: log likelihood = -1879.7733

```

Multinomial logistic regression

```

Number of obs   =      2048
LR chi2(36)     =      249.73
Prob > chi2     =      0.0000
Pseudo R2      =      0.0623

```

Log likelihood = -1879.7733

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.000904	.0088376	0.10	0.918	.9837317	1.018376
	exper	1.011993	.0120944	1.00	0.319	.9885635	1.035977
pyperstudlep		.9752532	.0070375	-3.47	0.001	.961557	.9891445
pyperstude~s		1.01315	.0057775	2.29	0.022	1.00189	1.024538
pyperstudw~e		.9357022	.0421799	-1.47	0.140	.8565779	1.022135
pyperstudb~k		.9308439	.0440153	-1.52	0.130	.8484523	1.021236
pyperstudl~o		.9267641	.0421093	-1.67	0.094	.8477997	1.013083
pyperstudm~e		.9854603	.0056638	-2.55	0.011	.9744218	.9966238
	male	1.73612	.2591064	3.70	0.000	1.295814	2.326039
	age	.9712952	.0101429	-2.79	0.005	.9516176	.9913797
	Asian	1.02e-06	.0010094	-0.01	0.989	0	.
	Black	1.343932	.4322819	0.92	0.358	.7154584	2.52447
	Latino	1.744471	.3064782	3.17	0.002	1.236289	2.461543
	Other	1.12765	1.309889	0.10	0.918	.1157203	10.9885
	AsianMale	2.69e+13	1.00e+17	0.01	0.993	0	.
	BlackMale	.9641619	.4580753	-0.08	0.939	.3799639	2.44657
	LatinoMale	.914876	.2283947	-0.36	0.722	.5608718	1.492316
	OtherMale	.6511142	1.070285	-0.26	0.794	.0259701	16.32451
3							
	tenure	1.035706	.0092516	3.93	0.000	1.017731	1.053999
	exper	1.077713	.0158415	5.09	0.000	1.047107	1.109213
pyperstudlep		1.009721	.0060714	1.61	0.108	.9978912	1.021691
pyperstude~s		1.042997	.0064487	6.81	0.000	1.030434	1.055713
pyperstudw~e		1.114564	.0615577	1.96	0.050	1.000214	1.241987
pyperstudb~k		1.068377	.061698	1.15	0.252	.9540441	1.196413
pyperstudl~o		1.064353	.0591947	1.12	0.262	.954433	1.186932
pyperstudm~e		1.004377	.0047549	0.92	0.356	.9951006	1.01374
	male	1.542055	.2472444	2.70	0.007	1.12622	2.11143
	age	.9518982	.0123852	-3.79	0.000	.9279306	.9764848
	Asian	1.022331	1.177671	0.02	0.985	.1069184	9.775309
	Black	1.108162	.355257	0.32	0.749	.5911845	2.077224
	Latino	1.068972	.1953692	0.36	0.715	.7471324	1.529449
	Other	1.34e-06	.0015247	-0.01	0.990	0	.
	AsianMale	.794321	5554.843	-0.00	1.000	0	.
	BlackMale	1.025192	.5059568	0.05	0.960	.3896881	2.697078
	LatinoMale	.9869801	.2636709	-0.05	0.961	.5846698	1.66612
	OtherMale	1344685	1.53e+09	0.01	0.990	0	.

by pyschooltype

```

. bysort pyschooltype: mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age A
> sian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1998a==1,
rrr

```

```

-----> pyschooltype =

```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -350.10119
Iteration 1: log likelihood = -317.01174
Iteration 2: log likelihood = -316.23251
Iteration 3: log likelihood = -316.11336
Iteration 4: log likelihood = -316.09383
Iteration 5: log likelihood = -316.08945
Iteration 6: log likelihood = -316.08836
Iteration 7: log likelihood = -316.08815
Iteration 8: log likelihood = -316.0881
Iteration 9: log likelihood = -316.08809

```

```

Multinomial logistic regression
Log likelihood = -316.08809
Number of obs   =      323
LR chi2(34)     =      68.03
Prob > chi2     =      0.0005
Pseudo R2      =      0.0972

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.976387	.0238563	-0.98	0.328	.9307313	1.024282
	exper	1.025689	.0306379	0.85	0.396	.9673635	1.087531
pyperstudlep		.9978431	.0119865	-0.18	0.857	.9746244	1.021615
pyperstude~s		.9812822	.0112357	-1.65	0.099	.9595058	1.003553
pyperstudw~e		.8907664	.041184	-2.50	0.012	.8135966	.9752558
pyperstudb~k		.8945075	.0437632	-2.28	0.023	.8127174	.9845289
pyperstudl~o		.9085424	.0422479	-2.06	0.039	.8293994	.9952375
pyperstudm~e		1.006451	.0120032	0.54	0.590	.9831984	1.030254
	male	2.239325	.8258581	2.19	0.029	1.086916	4.613584
	age	.9490397	.0241616	-2.05	0.040	.9028459	.997597
	Asian	3.71e+07	1.15e+11	0.01	0.996	0	.
	Black	2.422131	1.649749	1.30	0.194	.6374293	9.203719
	Latino	1.614308	.8674048	0.89	0.373	.5631395	4.627608
	Other	4.10e-07	.0004844	-0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	.5394981	.5609684	-0.59	0.553	.0702932	4.140631
	LatinoMale	.480859	.3765438	-0.94	0.350	.1036267	2.23133
	OtherMale	4.11e+13	1.37e+17	0.01	0.992	0	.
3							
	tenure	1.021138	.0215835	0.99	0.322	.9796991	1.064329
	exper	1.057017	.0303394	1.93	0.053	.9991944	1.118185
pyperstudlep		1.003145	.0112326	0.28	0.779	.981369	1.025404
pyperstude~s		.9948898	.010873	-0.47	0.639	.9738056	1.01643
pyperstudw~e		.9496246	.0447705	-1.10	0.273	.8658081	1.041555
pyperstudb~k		.9340106	.0469519	-1.36	0.174	.8463746	1.030721
pyperstudl~o		.9503869	.0455726	-1.06	0.289	.8651351	1.04404
pyperstudm~e		.9944982	.0137509	-0.40	0.690	.967909	1.021818
	male	1.835413	.6248911	1.78	0.074	.9417392	3.577147
	age	.9752813	.0244893	-1.00	0.319	.9284452	1.02448
	Asian	3.061156	14338.96	0.00	1.000	0	.
	Black	.4823815	.4238206	-0.83	0.407	.0862034	2.699335
	Latino	1.84589	.8975943	1.26	0.207	.7116972	4.787585
	Other	1.08e-06	.0012189	-0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	2.392877	2.859932	0.73	0.465	.2299198	24.90372
	LatinoMale	.3012749	.2350392	-1.54	0.124	.0652979	1.390038
	OtherMale	530243.9	2.55e+09	0.00	0.998	0	.

```

-----> pyschooltype = Both Elem/Sec

```

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -657.8572  
 Iteration 1: log likelihood = -604.87383  
 Iteration 2: log likelihood = -598.6686  
 Iteration 3: log likelihood = -598.53312  
 Iteration 4: log likelihood = -598.50376  
 Iteration 5: log likelihood = -598.4972  
 Iteration 6: log likelihood = -598.49576  
 Iteration 7: log likelihood = -598.49554  
 Iteration 8: log likelihood = -598.49552

Multinomial logistic regression	Number of obs	=	687
	LR chi2(34)	=	118.72
	Prob > chi2	=	0.0000
Log likelihood = -598.49552	Pseudo R2	=	0.0902

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.018131	.0198924	0.92	0.358	.9798801	1.057876
	exper	.9667455	.0238096	-1.37	0.170	.9211879	1.014556
pyperstudlep		1.03149	.0121054	2.64	0.008	1.008035	1.055491
pyperstude~s		.9815044	.0078017	-2.35	0.019	.9663318	.9969153
pyperstudw~e		.9203994	.036216	-2.11	0.035	.8520855	.9941903
pyperstudb~k		.9421267	.0399139	-1.41	0.159	.8670568	1.023696
pyperstudl~o		.9246973	.0385948	-1.88	0.061	.8520643	1.003522
pyperstudm~e		.9958653	.004745	-0.87	0.385	.9866086	1.005209
	male	1.865345	.5800371	2.00	0.045	1.014085	3.431183
	age	.9837192	.0214669	-0.75	0.452	.9425319	1.026706
	Asian	1.233054	3062.639	0.00	1.000	0	.
	Black	1.478746	.8317106	0.70	0.487	.4910673	4.452932
	Latino	2.868402	2.139305	1.41	0.158	.6649651	12.37318
	Other	9.65e-07	.00077	-0.02	0.986	0	.
	AsianMale	(omitted)					
	BlackMale	2.005486	1.456337	0.96	0.338	.4831667	8.324196
	LatinoMale	.8781417	.783266	-0.15	0.884	.1528721	5.044302
	OtherMale	1.16e+07	9.29e+09	0.02	0.984	0	.
3							
	tenure	1.011557	.0118615	0.98	0.327	.9885741	1.035074
	exper	1.026848	.0177925	1.53	0.126	.9925608	1.062319
pyperstudlep		.9928341	.0103395	-0.69	0.490	.9727745	1.013307
pyperstude~s		1.011248	.0054972	2.06	0.040	1.00053	1.02208
pyperstudw~e		1.086139	.0444507	2.02	0.043	1.00242	1.17685
pyperstudb~k		1.087534	.0465748	1.96	0.050	.9999751	1.182759
pyperstudl~o		1.080752	.0455338	1.84	0.065	.9950929	1.173785
pyperstudm~e		.9977135	.0034559	-0.66	0.509	.9909629	1.00451
	male	1.443435	.2820795	1.88	0.060	.9841324	2.117099
	age	.9720132	.0157965	-1.75	0.081	.9415405	1.003472
	Asian	3488767	3.77e+09	0.01	0.989	0	.
	Black	.2603918	.1753921	-2.00	0.046	.0695479	.9749244
	Latino	.6937551	.4462051	-0.57	0.570	.1966689	2.447241
	Other	1.82e-06	.0012987	-0.02	0.985	0	.
	AsianMale	(omitted)					
	BlackMale	6.588649	5.378276	2.31	0.021	1.330349	32.63077
	LatinoMale	2.193963	1.624592	1.06	0.289	.5139597	9.365466
	OtherMale	3269330	2.33e+09	0.02	0.983	0	.

-----> pyschooltype = Elementary

Iteration 0: log likelihood = -8943.5202  
 Iteration 1: log likelihood = -8446.1878

```

Iteration 2:  log likelihood = -8440.4106
Iteration 3:  log likelihood = -8440.1597
Iteration 4:  log likelihood = -8440.1425
Iteration 5:  log likelihood = -8440.1385
Iteration 6:  log likelihood = -8440.1378
Iteration 7:  log likelihood = -8440.1377
Iteration 8:  log likelihood = -8440.1377

```

```

Multinomial logistic regression      Number of obs   =      8455
                                     LR chi2(36)        =    1006.76
                                     Prob > chi2         =      0.0000
Log likelihood = -8440.1377          Pseudo R2        =      0.0563

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9922569	.0048489	-1.59	0.112	.9827986	1.001806
	exper	.9930642	.0059449	-1.16	0.245	.9814806	1.004785
pyperstudlep		1.007963	.0023071	3.47	0.001	1.003451	1.012495
pyperstude~s		.9862789	.0023482	-5.80	0.000	.9816873	.990892
pyperstudw~e		.976711	.0071978	-3.20	0.001	.9627049	.9909208
pyperstudb~k		.9806881	.0077209	-2.48	0.013	.9656716	.9959381
pyperstudl~o		.9832338	.0073977	-2.25	0.025	.9688409	.9978405
pyperstudm~e		1.010068	.0039648	2.55	0.011	1.002327	1.017869
	male	1.673942	.156841	5.50	0.000	1.393114	2.011381
	age	.993931	.0048602	-1.24	0.213	.9844507	1.003503
	Asian	.8850366	.4775337	-0.23	0.821	.3073858	2.548231
	Black	1.255977	.144062	1.99	0.047	1.003109	1.57259
	Latino	1.045414	.1050087	0.44	0.658	.8585938	1.272884
	Other	.7077082	.3105363	-0.79	0.431	.2994707	1.672454
	AsianMale	5.441915	7.043114	1.31	0.191	.4306231	68.77114
	BlackMale	.8059508	.1768317	-0.98	0.325	.5242621	1.238992
	LatinoMale	.8747093	.1514446	-0.77	0.439	.6230016	1.228113
	OtherMale	9.68e-07	.000534	-0.03	0.980	0	.
3							
	tenure	1.008341	.0039019	2.15	0.032	1.000723	1.016018
	exper	1.052094	.0057564	9.28	0.000	1.040872	1.063437
pyperstudlep		1.00036	.0020279	0.18	0.859	.9963937	1.004343
pyperstude~s		1.004869	.0020937	2.33	0.020	1.000774	1.008981
pyperstudw~e		1.005911	.0074686	0.79	0.427	.9913792	1.020656
pyperstudb~k		.9946445	.0078311	-0.68	0.495	.9794135	1.010112
pyperstudl~o		.9976277	.0075845	-0.31	0.755	.9828726	1.012604
pyperstudm~e		.9987684	.0037221	-0.33	0.741	.9914997	1.00609
	male	2.8833	.2185236	13.97	0.000	2.485294	3.345044
	age	1.003037	.0047094	0.65	0.518	.993849	1.01231
	Asian	.5639392	.3373512	-0.96	0.338	.174598	1.821483
	Black	1.050227	.1120115	0.46	0.646	.8521156	1.294397
	Latino	1.183703	.1060916	1.88	0.060	.9930064	1.411102
	Other	.5008255	.2227847	-1.55	0.120	.2094316	1.197652
	AsianMale	5.355743	6.638451	1.35	0.176	.4717958	60.79745
	BlackMale	.5553851	.1092975	-2.99	0.003	.3776428	.8167839
	LatinoMale	.5237233	.0790207	-4.29	0.000	.3896461	.7039363
	OtherMale	.8207717	.6340233	-0.26	0.798	.1805896	3.73037

```

-----> pyschooltype = Middle

```

```

Iteration 0:  log likelihood = -5135.6405
Iteration 1:  log likelihood = -4874.6318
Iteration 2:  log likelihood = -4869.5082
Iteration 3:  log likelihood = -4869.1342
Iteration 4:  log likelihood = -4869.071

```

```

Iteration 5: log likelihood = -4869.057
Iteration 6: log likelihood = -4869.0535
Iteration 7: log likelihood = -4869.0528
Iteration 8: log likelihood = -4869.0526
Iteration 9: log likelihood = -4869.0526
Iteration 10: log likelihood = -4869.0526

```

```

Multinomial logistic regression
Log likelihood = -4869.0526
Number of obs   = 4918
LR chi2(36)    = 533.18
Prob > chi2    = 0.0000
Pseudo R2     = 0.0519

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.006337	.0055262	1.15	0.250	.995564	1.017227
	exper	1.003285	.0069499	0.47	0.636	.9897559	1.017
pyperstudlep		1.010892	.0037422	2.93	0.003	1.003584	1.018253
pyperstude~s		.9931246	.0030039	-2.28	0.023	.9872545	.9990296
pyperstudw~e		.9846692	.0093174	-1.63	0.103	.9665758	1.003101
pyperstudb~k		.9864068	.0100605	-1.34	0.180	.9668845	1.006323
pyperstudl~o		.9839065	.0097338	-1.64	0.101	.9650124	1.003171
pyperstudm~e		1.008574	.0045048	1.91	0.056	.9997834	1.017442
	male	1.530024	.1289636	5.05	0.000	1.297035	1.804866
	age	.9657328	.005643	-5.97	0.000	.9547358	.9768564
	Asian	.5588788	.6175944	-0.53	0.599	.0640751	4.87468
	Black	1.169795	.1658721	1.11	0.269	.8859577	1.544567
	Latino	2.040125	.2875413	5.06	0.000	1.547697	2.689228
	Other	.8114674	.5582683	-0.30	0.761	.2107011	3.125183
	AsianMale	5.57e-07	.0004764	-0.02	0.987	0	.
	BlackMale	1.699391	.3410425	2.64	0.008	1.146753	2.518354
	LatinoMale	1.011672	.1822931	0.06	0.949	.7106595	1.440182
	OtherMale	1.649121	1.469579	0.56	0.575	.287556	9.457634
3							
	tenure	1.003515	.0054959	0.64	0.522	.9928006	1.014345
	exper	1.060257	.0087274	7.11	0.000	1.043289	1.077501
pyperstudlep		1.001882	.0044537	0.42	0.672	.9931911	1.01065
pyperstude~s		1.011006	.0034869	3.17	0.002	1.004195	1.017864
pyperstudw~e		1.037255	.0133719	2.84	0.005	1.011375	1.063797
pyperstudb~k		1.019422	.0140396	1.40	0.162	.9922731	1.047314
pyperstudl~o		1.019301	.0136523	1.43	0.153	.9928911	1.046413
pyperstudm~e		.9999435	.0056445	-0.01	0.992	.9889415	1.011068
	male	2.260264	.2071175	8.90	0.000	1.888687	2.704945
	age	.9682615	.0071784	-4.35	0.000	.9542938	.9824336
	Asian	4.31e-07	.0005636	-0.01	0.991	0	.
	Black	.978628	.1873952	-0.11	0.910	.6723927	1.424335
	Latino	1.988037	.3425668	3.99	0.000	1.418242	2.786753
	Other	.9399014	.7511548	-0.08	0.938	.1962524	4.50142
	AsianMale	2091090	2.74e+09	0.01	0.991	0	.
	BlackMale	1.754345	.4343105	2.27	0.023	1.079914	2.849974
	LatinoMale	.7852549	.1653762	-1.15	0.251	.5196919	1.186521
	OtherMale	1.180172	1.17851	0.17	0.868	.1667027	8.355032

```

-----> pyschooltype = Secondary

```

```

Iteration 0: log likelihood = -6053.7907
Iteration 1: log likelihood = -5753.7091
Iteration 2: log likelihood = -5740.5649
Iteration 3: log likelihood = -5740.3551
Iteration 4: log likelihood = -5740.3107
Iteration 5: log likelihood = -5740.3004

```



```

Iteration 6: log likelihood = -5740.298
Iteration 7: log likelihood = -5740.2975
Iteration 8: log likelihood = -5740.2974
Iteration 9: log likelihood = -5740.2974

```

```

Multinomial logistic regression
Log likelihood = -5740.2974

Number of obs   =      6392
LR chi2(36)     =      626.99
Prob > chi2     =      0.0000
Pseudo R2      =      0.0518

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9978515	.0043409	-0.49	0.621	.9893797	1.006396
	exper	.9996235	.0058122	-0.06	0.948	.9882965	1.011108
pyperstudlep		1.011217	.0038386	2.94	0.003	1.003722	1.018769
pyperstude~s		.9951173	.0023484	-2.07	0.038	.9905251	.9997308
pyperstudw~e		.9810787	.0074644	-2.51	0.012	.9665572	.9958183
pyperstudb~k		.9842037	.0078794	-1.99	0.047	.9688809	.9997687
pyperstudl~o		.980294	.0076767	-2.54	0.011	.9653628	.9954562
pyperstudm~e		.9996487	.0026152	-0.13	0.893	.9945362	1.004787
	male	1.283617	.0955315	3.35	0.001	1.109394	1.4852
	age	.9808962	.005279	-3.58	0.000	.9706039	.9912977
	Asian	1.549011	1.078439	0.63	0.530	.3957662	6.062762
	Black	1.854265	.2585743	4.43	0.000	1.410825	2.437084
	Latino	1.813825	.256455	4.21	0.000	1.374817	2.393018
	Other	.7571792	.4248201	-0.50	0.620	.2521313	2.273896
	AsianMale	1.461202	1.308913	0.42	0.672	.2524772	8.45665
	BlackMale	1.426686	.2631938	1.93	0.054	.9938007	2.048129
	LatinoMale	1.45167	.2436385	2.22	0.026	1.044739	2.017102
	OtherMale	2.324188	1.626551	1.21	0.228	.5896168	9.161631
3							
	tenure	.9783483	.0048513	-4.41	0.000	.9688859	.987903
	exper	1.052563	.007999	6.74	0.000	1.037002	1.068358
pyperstudlep		.9982822	.0051023	-0.34	0.737	.9883317	1.008333
pyperstude~s		1.019285	.0029103	6.69	0.000	1.013597	1.025005
pyperstudw~e		1.052808	.0144824	3.74	0.000	1.024802	1.081579
pyperstudb~k		1.035584	.0148568	2.44	0.015	1.00687	1.065116
pyperstudl~o		1.032081	.0144883	2.25	0.024	1.004071	1.060872
pyperstudm~e		1.018841	.002543	7.48	0.000	1.013869	1.023837
	male	2.308505	.2230171	8.66	0.000	1.910289	2.789733
	age	.9829839	.0073253	-2.30	0.021	.9687309	.9974466
	Asian	1.38e-06	.0010794	-0.02	0.986	0	.
	Black	1.278673	.2982891	1.05	0.292	.8094515	2.019891
	Latino	1.420314	.3000277	1.66	0.097	.9388066	2.148783
	Other	.8002782	.6259356	-0.28	0.776	.172773	3.70686
	AsianMale	1701556	1.33e+09	0.02	0.985	0	.
	BlackMale	.8863349	.263988	-0.41	0.685	.494395	1.588992
	LatinoMale	1.059658	.2550397	0.24	0.810	.6611466	1.698376
	OtherMale	1.39135	1.313959	0.35	0.727	.2185652	8.857099

### 1998-99

```

. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino Oth
> er AsianMale BlackMale LatinoMale OtherMale if validcert1999a==1, rrr

```

```

Iteration 0: log likelihood = -22741.443
Iteration 1: log likelihood = -21704.446
Iteration 2: log likelihood = -21685.121
Iteration 3: log likelihood = -21685.091
Iteration 4: log likelihood = -21685.091

```

Multinomial logistic regression

Number of obs = 21713  
 LR chi2(36) = 2112.70  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0465

Log likelihood = -21685.091

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.002158	.0026402	0.82	0.413	.9969965	1.007346
exper	.99396	.0033194	-1.81	0.070	.9874753	1.000487
pyperstudlep	1.00909	.0015065	6.06	0.000	1.006142	1.012047
pyperstude~s	.9945443	.0012042	-4.52	0.000	.9921869	.9969073
pyperstudw~e	.9734326	.0042597	-6.15	0.000	.9651194	.9818175
pyperstudb~k	.975292	.0045566	-5.35	0.000	.9664019	.9842638
pyperstudl~o	.9747438	.0043667	-5.71	0.000	.9662226	.9833401
pyperstudm~e	.9977051	.0014714	-1.56	0.119	.9948253	1.000593
male	1.293825	.0550899	6.05	0.000	1.190233	1.406432
age	.984028	.0028078	-5.64	0.000	.9785401	.9895467
Asian	1.123528	.3997356	0.33	0.743	.5594206	2.25647
Black	1.445661	.102137	5.22	0.000	1.258718	1.660368
Latino	1.371846	.0899925	4.82	0.000	1.206332	1.560069
Other	.7351951	.2185579	-1.03	0.301	.4105428	1.316579
AsianMale	1.040541	.5619399	0.07	0.941	.3610535	2.998797
BlackMale	1.54516	.1620242	4.15	0.000	1.258106	1.89771
LatinoMale	1.271583	.1136908	2.69	0.007	1.067186	1.51513
OtherMale	1.413313	.5753701	0.85	0.395	.6363672	3.138837
3						
tenure	1.003038	.0023707	1.28	0.199	.9984023	1.007695
exper	1.053025	.0037237	14.61	0.000	1.045752	1.060348
pyperstudlep	1.012108	.0015101	8.07	0.000	1.009153	1.015072
pyperstude~s	1.022798	.0012721	18.12	0.000	1.020308	1.025294
pyperstudw~e	1.01501	.0052617	2.87	0.004	1.004749	1.025375
pyperstudb~k	.9937519	.0054981	-1.13	0.257	.9830341	1.004586
pyperstudl~o	.9892349	.0052683	-2.03	0.042	.978963	.9996146
pyperstudm~e	1.002256	.001466	1.54	0.123	.9993874	1.005134
male	1.458772	.0594008	9.27	0.000	1.346873	1.579967
age	.9907014	.0031663	-2.92	0.003	.9845149	.9969267
Asian	.508103	.2385526	-1.44	0.149	.2024489	1.275228
Black	.9134977	.0725799	-1.14	0.255	.7817667	1.067426
Latino	1.26053	.0858313	3.40	0.001	1.103046	1.440498
Other	.5451471	.1803445	-1.83	0.067	.2850498	1.042573
AsianMale	2.13523	1.321009	1.23	0.220	.6350811	7.178935
BlackMale	1.317799	.1586878	2.29	0.022	1.040755	1.66859
LatinoMale	1.010024	.0945811	0.11	0.915	.8406653	1.213501
OtherMale	1.77502	.7720091	1.32	0.187	.7568158	4.163093

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=21713)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	1.813	2	0.404
exper	270.294	2	0.000
pyperstudlep	77.015	2	0.000
pyperstude~s	459.348	2	0.000
pyperstudw~e	64.854	2	0.000
pyperstudb~k	28.839	2	0.000
pyperstudl~o	32.411	2	0.000
pyperstudm~e	6.801	2	0.033

male		97.105	2	0.000
age		34.036	2	0.000
Asian		2.891	2	0.236
Black		34.605	2	0.000
Latino		27.025	2	0.000
Other		4.092	2	0.129
AsianMale		1.637	2	0.441
BlackMale		18.032	2	0.000
LatinoMale		8.136	2	0.017
OtherMale		2.058	2	0.357

\*\*\*\* Wald tests for independent variables (N=21713)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		1.813	2	0.404
exper		254.304	2	0.000
pyperstudlep		75.960	2	0.000
pyperstude~s		437.272	2	0.000
pyperstudw~e		65.456	2	0.000
pyperstudb~k		29.143	2	0.000
pyperstudl~o		32.623	2	0.000
pyperstudm~e		6.780	2	0.034
male		96.805	2	0.000
age		33.739	2	0.000
Asian		2.527	2	0.283
Black		35.263	2	0.000
Latino		27.055	2	0.000
Other		3.831	2	0.147
AsianMale		1.584	2	0.453
BlackMale		17.884	2	0.000
LatinoMale		8.139	2	0.017
OtherMale		2.025	2	0.363

\*\*\*\* Hausman tests of IIA assumption (N=21713)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-115.910	18	---	---
3		-11.867	18	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=21713)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-4883.874	-4876.829	14.090	19	0.778	for Ho
3		-4982.943	-4971.622	22.642	19	0.253	for Ho

\*\*\*\* Wald tests for combining alternatives (N=21713)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
---------------------	--	------	----	--------

```

2-      3 | 1374.163   18   0.000
2-      1 |  491.770   18   0.000
3-      1 | 1161.536   18   0.000
-----

**** LR tests for combining alternatives (N=21713)

Ho: All coefficients except intercepts associated with a given pair
    of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested|      chi2    df    P>chi2
-----+-----
2-      3 | 1538.047   18   0.000
2-      1 |  503.417   18   0.000
3-      1 | 1281.840   18   0.000
-----

by region
. bysort region: mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1999a==1, rrr

-----> region = 1

note: AsianMale omitted because of collinearity
Iteration 0:  log likelihood = -1566.0138
Iteration 1:  log likelihood = -1474.5074
Iteration 2:  log likelihood = -1470.9544
Iteration 3:  log likelihood = -1470.8095
Iteration 4:  log likelihood = -1470.7763
Iteration 5:  log likelihood = -1470.769
Iteration 6:  log likelihood = -1470.7674
Iteration 7:  log likelihood = -1470.7671
Iteration 8:  log likelihood = -1470.7671

Multinomial logistic regression                                Number of obs   =       1472
                                                             LR chi2(34)     =       190.49
                                                             Prob > chi2     =       0.0000
Log likelihood = -1470.7671                                Pseudo R2      =       0.0608

-----
      admin |          RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1          | (base outcome)
-----+-----
2          |
   tenure |   1.000981   .0101292     0.10   0.923     .9813234   1.021032
   exper  |   1.023036   .0134918     1.73   0.084     .9969317   1.049825
pyperstudlep | 1.000242   .0035467     0.07   0.946     .9933146   1.007217
pyperstude~s | 1.000687   .0072536     0.09   0.924     .986571   1.015006
pyperstudw~e | 1.096336   .1213151     0.83   0.406     .8825796   1.361863
pyperstudb~k | 1.177899   .16262      1.19   0.236     .8986522   1.543919
pyperstudl~o | 1.095153   .1156105     0.86   0.389     .8904659   1.346891
pyperstudm~e | .9835536   .0078765    -2.07   0.038     .9682365   .999113
   male   | 1.224928   .3268214     0.76   0.447     .7261106   2.06642
   age    | .9802001   .0110308    -1.78   0.076     .9588167   1.00206
   Asian  | 9.41e-07   .000916     -0.01   0.989      0         .
   Black  | 5958488   9.90e+09     0.01   0.993      0         .
   Latino | 1.049218   .2198983     0.23   0.819     .6957747   1.582207
   Other  | 1.948232   2.786977     0.47   0.641     .1180275   32.15866
   AsianMale | (omitted)
   BlackMale | 1.81e-13   3.90e-10    -0.01   0.989      0         .
   LatinoMale | 1.619562   .4901268     1.59   0.111     .894951   2.930867
   OtherMale | 1.698186   3.223217     0.28   0.780     .0411505   70.0803
-----
3          |
   tenure | 1.004339   .010247     0.42   0.671     .984455   1.024625

```

exper		1.109524	.0190675	6.05	0.000	1.072775	1.147532
pyperstudlep		1.014772	.0037939	3.92	0.000	1.007364	1.022236
pyperstude~s		1.020381	.0087988	2.34	0.019	1.003281	1.037773
pyperstudw~e		.9833652	.1232187	-0.13	0.894	.7692308	1.257109
pyperstudb~k		.8961236	.2116604	-0.46	0.642	.5640504	1.423698
pyperstudl~o		.9445988	.1124651	-0.48	0.632	.7480011	1.192868
pyperstudm~e		1.007036	.0067529	1.05	0.296	.9938868	1.020358
male		.6493551	.2220571	-1.26	0.207	.3322008	1.269299
age		.9542397	.014977	-2.98	0.003	.9253322	.9840503
Asian		3.14e-07	.0003051	-0.02	0.988	0	.
Black		4.49262	12597.58	0.00	1.000	0	.
Latino		1.125385	.2549183	0.52	0.602	.721921	1.754334
Other		3.24e-06	.0025643	-0.02	0.987	0	.
AsianMale		(omitted)					
BlackMale		1.53e-07	.0004861	-0.00	0.996	0	.
LatinoMale		2.516415	.9442712	2.46	0.014	1.206073	5.250384
OtherMale		1024474	8.12e+08	0.02	0.986	0	.

-----> region = 2

```

Iteration 0: log likelihood = -820.32948
Iteration 1: log likelihood = -772.85465
Iteration 2: log likelihood = -771.01487
Iteration 3: log likelihood = -770.85904
Iteration 4: log likelihood = -770.83261
Iteration 5: log likelihood = -770.82672
Iteration 6: log likelihood = -770.82527
Iteration 7: log likelihood = -770.82497
Iteration 8: log likelihood = -770.82491
Iteration 9: log likelihood = -770.8249

```

Multinomial logistic regression	Number of obs	=	853
	LR chi2(36)	=	99.01
	Prob > chi2	=	0.0000
Log likelihood = -770.8249	Pseudo R2	=	0.0603

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9919595	.0148799	-0.54	0.590	.96322 1.021556
exper		.9767895	.0177548	-1.29	0.196	.9426032 1.012216
pyperstudlep		1.000357	.0170567	0.02	0.983	.9674787 1.034352
pyperstude~s		1.008631	.0066396	1.31	0.192	.9957008 1.021728
pyperstudw~e		.8932376	.0578367	-1.74	0.081	.7867777 1.014103
pyperstudb~k		.908587	.0590439	-1.48	0.140	.7999296 1.032004
pyperstudl~o		.8904336	.054505	-1.90	0.058	.7897653 1.003934
pyperstudm~e		.9827891	.0094993	-1.80	0.072	.964346 1.001585
male		1.182127	.3170049	0.62	0.533	.6988783 1.999523
age		1.000958	.0153039	0.06	0.950	.9714079 1.031407
Asian		5.71e+07	2.35e+11	0.00	0.997	0 .
Black		.9830542	.8344692	-0.02	0.984	.1862224 5.189471
Latino		.9736417	.259545	-0.10	0.920	.5774218 1.641743
Other		1.27e-06	.0018617	-0.01	0.993	0 .
AsianMale		1.27e-14	5.90e-11	-0.01	0.995	0 .
BlackMale		3.136485	3.507347	1.02	0.307	.3504247 28.07319
LatinoMale		1.410695	.552739	0.88	0.380	.654508 3.040544
OtherMale		1343879	1.97e+09	0.01	0.992	0 .
3						
tenure		.9950369	.0119706	-0.41	0.679	.9718494 1.018778
exper		1.065573	.0199311	3.40	0.001	1.027217 1.105362
pyperstudlep		1.003726	.0133851	0.28	0.780	.9778314 1.030306

pyperstude~s		1.030196	.0066843	4.58	0.000	1.017178	1.043381
pyperstudw~e		1.173431	.0886834	2.12	0.034	1.011875	1.36078
pyperstudb~k		1.121005	.0883974	1.45	0.147	.9604743	1.308367
pyperstudl~o		1.141481	.0829028	1.82	0.068	.9900291	1.316101
pyperstudm~e		.991919	.0074985	-1.07	0.283	.9773305	1.006725
male		1.152568	.3003827	0.54	0.586	.6915543	1.92091
age		.9846156	.0168671	-0.91	0.365	.9521055	1.018236
Asian		1.742185	13428.37	0.00	1.000	0	.
Black		1.579775	1.092704	0.66	0.509	.4072241	6.128541
Latino		.9504612	.2354255	-0.21	0.837	.5849182	1.54445
Other		3.306121	4.772157	0.83	0.407	.1952848	55.97178
AsianMale		1.17e-06	.0092686	-0.00	0.999	0	.
BlackMale		2.791156	2.772671	1.03	0.301	.3983088	19.55908
LatinoMale		1.496739	.5433461	1.11	0.267	.7347554	3.048942
OtherMale		1.49e-06	.0012303	-0.02	0.987	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -331.98121  
Iteration 1: log likelihood = -296.22361  
Iteration 2: log likelihood = -293.82957  
Iteration 3: log likelihood = -293.70684  
Iteration 4: log likelihood = -293.67988  
Iteration 5: log likelihood = -293.67352  
Iteration 6: log likelihood = -293.67221  
Iteration 7: log likelihood = -293.67199  
Iteration 8: log likelihood = -293.67197

Multinomial logistic regression	Number of obs	=	319
	LR chi2(30)	=	76.62
	Prob > chi2	=	0.0000
Log likelihood = -293.67197	Pseudo R2	=	0.1154

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.021886	.0242185	0.91	0.361	.9755043 1.070473
exper		.9553591	.0352851	-1.24	0.216	.8886453 1.027081
pyperstudlep		.9889923	.0511771	-0.21	0.831	.8936059 1.094561
pyperstude~s		.9817473	.0142238	-1.27	0.204	.9542613 1.010025
pyperstudw~e		1.040024	.0587611	0.69	0.487	.9310022 1.161812
pyperstudb~k		1.047095	.0599828	0.80	0.422	.9358904 1.171513
pyperstudl~o		1.06558	.060948	1.11	0.267	.9525761 1.191989
pyperstudm~e		.9886988	.0172698	-0.65	0.515	.9554234 1.023133
male		.7559743	.3021378	-0.70	0.484	.3453903 1.654642
age		.9711985	.0309602	-0.92	0.359	.9123745 1.033815
Asian		(omitted)				
Black		1.174158	.9159293	0.21	0.837	.2545232 5.416591
Latino		2.375611	1.396151	1.47	0.141	.750797 7.516717
Other		1528115	1.01e+09	0.02	0.983	0 .
AsianMale		(omitted)				
BlackMale		2.149158	3.665789	0.45	0.654	.0759252 60.83464
LatinoMale		1.238207	1.213556	0.22	0.827	.1813587 8.453728
OtherMale		(omitted)				
-----						
3						
tenure		.9833639	.0169006	-0.98	0.329	.950791 1.017053
exper		1.016555	.0298211	0.56	0.576	.9597556 1.076717
pyperstudlep		1.078844	.0423078	1.94	0.053	.999029 1.165036

pyperstude~s		1.029236	.012136	2.44	0.015	1.005723	1.053299
pyperstudw~e		1.100844	.0472677	2.24	0.025	1.011992	1.197497
pyperstudb~k		1.080676	.0463563	1.81	0.070	.9935334	1.175461
pyperstudl~o		1.07159	.046089	1.61	0.108	.9849601	1.16584
pyperstudm~e		1.005992	.0123038	0.49	0.625	.9821638	1.030398
male		2.661355	.8305438	3.14	0.002	1.443656	4.906163
age		1.014616	.0282495	0.52	0.602	.960732	1.071523
Asian		(omitted)					
Black		1.68508	1.34638	0.65	0.514	.3519734	8.067354
Latino		.574855	.4987902	-0.64	0.523	.1049513	3.148683
Other		641307.7	4.24e+08	0.02	0.984	0	.
AsianMale		(omitted)					
BlackMale		1.39e-06	.0009981	-0.02	0.985	0	.
LatinoMale		4.76908	5.143269	1.45	0.147	.5760319	39.48414
OtherMale		(omitted)					

-----> region = 4

Iteration 0: log likelihood = -4560.8094  
Iteration 1: log likelihood = -4291.7573  
Iteration 2: log likelihood = -4280.6977  
Iteration 3: log likelihood = -4280.6699  
Iteration 4: log likelihood = -4280.6699

Multinomial logistic regression	Number of obs	=	4412
	LR chi2(36)	=	560.28
	Prob > chi2	=	0.0000
Log likelihood = -4280.6699	Pseudo R2	=	0.0614

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.01171	.0058507	2.01	0.044	1.000308 1.023242
exper		1.030474	.0084378	3.67	0.000	1.014068 1.047145
pyperstudlep		.9996599	.0041022	-0.08	0.934	.9916519 1.007733
pyperstude~s		.9999993	.0027218	-0.00	1.000	.9946789 1.005348
pyperstudw~e		.9860255	.0065405	-2.12	0.034	.9732893 .9989283
pyperstudb~k		.9792735	.0071751	-2.86	0.004	.9653111 .9934378
pyperstudl~o		.9868483	.0069243	-1.89	0.059	.9733698 1.000513
pyperstudm~e		1.000681	.0031011	0.22	0.826	.9946211 1.006777
male		1.147494	.1065007	1.48	0.138	.9566412 1.376422
age		.956743	.0065764	-6.43	0.000	.9439399 .9697197
Asian		.4952527	.3300985	-1.05	0.292	.1341157 1.828834
Black		1.011383	.116823	0.10	0.922	.8064823 1.268342
Latino		1.482259	.246215	2.37	0.018	1.070368 2.052653
Other		.5406066	.3601655	-0.92	0.356	.1464839 1.995138
AsianMale		.4982208	.6315766	-0.55	0.583	.0415321 5.976673
BlackMale		1.834011	.3198658	3.48	0.001	1.303004 2.581417
LatinoMale		1.146066	.304442	0.51	0.608	.6809189 1.928961
OtherMale		2.076589	2.371463	0.64	0.522	.2214516 19.47252
3						
tenure		1.02895	.00653	4.50	0.000	1.016231 1.041829
exper		1.095911	.0113848	8.82	0.000	1.073823 1.118454
pyperstudlep		1.019274	.0049713	3.91	0.000	1.009577 1.029064
pyperstude~s		1.022823	.0036548	6.32	0.000	1.015684 1.030011
pyperstudw~e		1.010308	.0087501	1.18	0.236	.9933026 1.027604
pyperstudb~k		.9911001	.0097476	-0.91	0.363	.9721782 1.01039
pyperstudl~o		.9787975	.0091691	-2.29	0.022	.9609903 .9969346
pyperstudm~e		1.001544	.0039564	0.39	0.696	.9938191 1.009328
male		1.120359	.1212374	1.05	0.294	.9062463 1.385059
age		.9633458	.008713	-4.13	0.000	.946419 .9805753

Asian		.2150155	.2353961	-1.40	0.160	.0251528	1.838036
Black		.4906748	.0746234	-4.68	0.000	.3642002	.6610698
Latino		1.667455	.3102393	2.75	0.006	1.157933	2.401181
Other		1.198844	.7477448	0.29	0.771	.3530614	4.070759
AsianMale		1.253122	1.950014	0.15	0.885	.0593496	26.45874
BlackMale		1.913179	.4358373	2.85	0.004	1.22418	2.989965
LatinoMale		1.060235	.3321345	0.19	0.852	.5737872	1.959087
OtherMale		1.102407	1.473569	0.07	0.942	.0802705	15.14009

-----> region = 5

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -556.27594  
Iteration 1: log likelihood = -493.0266  
Iteration 2: log likelihood = -490.52079  
Iteration 3: log likelihood = -490.33586  
Iteration 4: log likelihood = -490.3023  
Iteration 5: log likelihood = -490.29497  
Iteration 6: log likelihood = -490.29324  
Iteration 7: log likelihood = -490.29285  
Iteration 8: log likelihood = -490.29277  
Iteration 9: log likelihood = -490.29275

Multinomial logistic regression	Number of obs	=	559
	LR chi2(32)	=	131.97
	Prob > chi2	=	0.0000
Log likelihood = -490.29275	Pseudo R2	=	0.1186

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.015005	.0184129	0.82	0.412	.97955 1.051742
exper		1.004536	.0270558	0.17	0.867	.9528827 1.058989
pyperstudlep		.9514335	.0440217	-1.08	0.282	.8689491 1.041748
pyperstude~s		1.008982	.0089172	1.01	0.312	.9916547 1.026611
pyperstudw~e		.9615706	.0406757	-0.93	0.354	.885063 1.044692
pyperstudb~k		.9522405	.0412482	-1.13	0.259	.8747323 1.036617
pyperstudl~o		.9917606	.0570815	-0.14	0.886	.8859625 1.110193
pyperstudm~e		1.013714	.0115978	1.19	0.234	.991236 1.036702
male		2.831736	.8012679	3.68	0.000	1.626286 4.930702
age		.9416364	.0209544	-2.70	0.007	.9014492 .9836151
Asian		(omitted)				
Black		1.627253	.6582145	1.20	0.229	.7364593 3.595516
Latino		3.71234	4.80665	1.01	0.311	.2934485 46.96382
Other		2.89e-07	.0007619	-0.01	0.995	0 .
AsianMale		(omitted)				
BlackMale		1.332537	.6861903	0.56	0.577	.4856832 3.655997
LatinoMale		1.44e-07	.0001907	-0.01	0.991	0 .
OtherMale		.8379068	3126.325	-0.00	1.000	0 .
-----						
3						
tenure		1.000309	.0153708	0.02	0.984	.9706321 1.030894
exper		1.104063	.0307493	3.55	0.000	1.045411 1.166006
pyperstudlep		1.072958	.0551981	1.37	0.171	.9700474 1.186787
pyperstude~s		1.03408	.0090612	3.82	0.000	1.016472 1.051992
pyperstudw~e		1.085205	.0479475	1.85	0.064	.9951841 1.18337
pyperstudb~k		1.050771	.0476249	1.09	0.275	.9614542 1.148386
pyperstudl~o		1.050545	.0547491	0.95	0.344	.9485373 1.163523
pyperstudm~e		1.019536	.0111658	1.77	0.077	.9978845 1.041657
male		3.146429	.8319919	4.33	0.000	1.873871 5.28319
age		.9473188	.0233908	-2.19	0.028	.9025654 .9942913



Asian		(omitted)					
Black		1.260121	.5605067	0.52	0.603	.5269804	3.013214
Latino		2.33e-06	.0029464	-0.01	0.992	0	.
Other		7.03e-07	.0017468	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		.8566655	.5078494	-0.26	0.794	.2680409	2.737925
LatinoMale		318051	4.02e+08	0.01	0.992	0	.
OtherMale		.7672296	2695.068	-0.00	1.000	0	.

-----> region = 6

Iteration 0: log likelihood = -789.8023  
Iteration 1: log likelihood = -742.52844  
Iteration 2: log likelihood = -741.51663  
Iteration 3: log likelihood = -741.37029  
Iteration 4: log likelihood = -741.34796  
Iteration 5: log likelihood = -741.34255  
Iteration 6: log likelihood = -741.34143  
Iteration 7: log likelihood = -741.3412  
Iteration 8: log likelihood = -741.34114  
Iteration 9: log likelihood = -741.34113

Multinomial logistic regression	Number of obs	=	750
	LR chi2(36)	=	96.92
	Prob > chi2	=	0.0000
Log likelihood = -741.34113	Pseudo R2	=	0.0614

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9685549	.0142399	-2.17	0.030	.9410435 .9968706
exper		.9862204	.0182973	-0.75	0.455	.9510027 1.022742
pyperstudlep		1.015481	.0197252	0.79	0.429	.9775467 1.054887
pyperstude~s		1.002809	.0077509	0.36	0.717	.9877324 1.018116
pyperstudw~e		.9599686	.0442087	-0.89	0.375	.8771166 1.050647
pyperstudb~k		.955604	.045607	-0.95	0.341	.8702692 1.049306
pyperstudl~o		.9612424	.0457372	-0.83	0.406	.8756521 1.055199
pyperstudm~e		1.000624	.0082702	0.08	0.940	.9845457 1.016966
male		1.097809	.2250133	0.46	0.649	.7346148 1.640567
age		.9795067	.0161205	-1.26	0.208	.9484153 1.011617
Asian		9.70e-07	.0017564	-0.01	0.994	0 .
Black		.9051543	.5174018	-0.17	0.862	.2952304 2.775135
Latino		.1699565	.1872848	-1.61	0.108	.0196042 1.473417
Other		9.05e-07	.00115	-0.01	0.991	0 .
AsianMale		1307727	5.02e+09	0.00	0.997	0 .
BlackMale		2.320514	1.871707	1.04	0.297	.4775426 11.27603
LatinoMale		6.052516	8.229589	1.32	0.185	.421269 86.95856
OtherMale		.8489423	1878.497	-0.00	1.000	0 .
-----						
3						
tenure		.978189	.012373	-1.74	0.081	.9542364 1.002743
exper		1.061861	.0206515	3.09	0.002	1.022146 1.103118
pyperstudlep		1.026881	.0207974	1.31	0.190	.9869176 1.068463
pyperstude~s		1.031248	.007398	4.29	0.000	1.01685 1.045851
pyperstudw~e		.9472791	.0412008	-1.25	0.213	.8698732 1.031573
pyperstudb~k		.9422966	.0419803	-1.33	0.182	.8635067 1.028276
pyperstudl~o		.9094198	.0405509	-2.13	0.033	.8333156 .9924744
pyperstudm~e		1.005569	.0076629	0.73	0.466	.9906611 1.0207
male		1.600209	.3142748	2.39	0.017	1.08894 2.351524
age		.9662461	.0173659	-1.91	0.056	.932802 1.000889
Asian		1.35e-06	.0022771	-0.01	0.994	0 .
Black		.977875	.537736	-0.04	0.968	.3328169 2.87317

Latino		.691973	.5737454	-0.44	0.657	.1362449	3.514455
Other		1.42e-06	.0016906	-0.01	0.991	0	.
AsianMale		5.42e+11	1.40e+15	0.01	0.992	0	.
BlackMale		.8063999	.662459	-0.26	0.793	.161171	4.034726
LatinoMale		2.412492	2.668786	0.80	0.426	.2759537	21.09093
OtherMale		.3198658	660.4881	-0.00	1.000	0	.

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1035.3165  
Iteration 1: log likelihood = -938.04758  
Iteration 2: log likelihood = -932.63315  
Iteration 3: log likelihood = -932.50966  
Iteration 4: log likelihood = -932.48288  
Iteration 5: log likelihood = -932.47664  
Iteration 6: log likelihood = -932.47534  
Iteration 7: log likelihood = -932.47512  
Iteration 8: log likelihood = -932.4751

Multinomial logistic regression	Number of obs	=	1048
	LR chi2(34)	=	205.68
	Prob > chi2	=	0.0000
Log likelihood = -932.4751	Pseudo R2	=	0.0993

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9746926	.013335	-1.87	0.061	.9489038 1.001182
exper		1.013536	.0181176	0.75	0.452	.9786408 1.049675
pyperstudlep		1.0133	.0226552	0.59	0.555	.9698551 1.05869
pyperstude~s		.9830248	.0079283	-2.12	0.034	.9676078 .9986873
pyperstudw~e		.8576227	.1055485	-1.25	0.212	.6738109 1.091577
pyperstudb~k		.8652161	.1074117	-1.17	0.244	.6783484 1.103561
pyperstudl~o		.8630406	.1079575	-1.18	0.239	.6753901 1.102828
pyperstudm~e		.9808629	.0133415	-1.42	0.155	.9550594 1.007363
male		1.681127	.3620483	2.41	0.016	1.102263 2.563986
age		.9745888	.0153084	-1.64	0.101	.945042 1.005059
Asian		.8470303	2088.313	-0.00	1.000	0 .
Black		3.773319	1.246441	4.02	0.000	1.974903 7.209433
Latino		.8246459	.9491358	-0.17	0.867	.0864106 7.869875
Other		2.415178	3.018512	0.71	0.480	.2085016 27.9762
AsianMale		(omitted)				
BlackMale		1.374344	.6536677	0.67	0.504	.5410605 3.490963
LatinoMale		7.695801	11.10268	1.41	0.157	.4552303 130.0998
OtherMale		1.161337	2.208254	0.08	0.937	.0279524 48.25
3						
tenure		.9754335	.0100145	-2.42	0.015	.9560016 .9952602
exper		1.091221	.0174612	5.46	0.000	1.057529 1.125986
pyperstudlep		1.036571	.0204925	1.82	0.069	.9971742 1.077523
pyperstude~s		1.039846	.0063307	6.42	0.000	1.027512 1.052328
pyperstudw~e		.832949	.0947522	-1.61	0.108	.6664843 1.040991
pyperstudb~k		.8115185	.0929651	-1.82	0.068	.6483169 1.015803
pyperstudl~o		.7898595	.0915288	-2.04	0.042	.6293797 .9912584
pyperstudm~e		1.013292	.0082156	1.63	0.103	.9973168 1.029523
male		1.713802	.2869647	3.22	0.001	1.234334 2.379516
age		.960373	.0139477	-2.78	0.005	.9334214 .9881028
Asian		1056722	1.25e+09	0.01	0.991	0 .
Black		1.03251	.3497172	0.09	0.925	.5316025 2.005401
Latino		.4103019	.4686606	-0.78	0.435	.0437353 3.849236
Other		3.59e-06	.001982	-0.02	0.982	0 .

AsianMale		(omitted)					
BlackMale		2.162922	1.06375	1.57	0.117	.8249169	5.671156
LatinoMale		5.05078	7.779503	1.05	0.293	.2467678	103.378
OtherMale		1645788	9.08e+08	0.03	0.979	0	.

-----> region = 8

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -324.79286  
Iteration 1: log likelihood = -290.33052  
Iteration 2: log likelihood = -288.4901  
Iteration 3: log likelihood = -288.26221  
Iteration 4: log likelihood = -288.22246  
Iteration 5: log likelihood = -288.21355  
Iteration 6: log likelihood = -288.21141  
Iteration 7: log likelihood = -288.21096  
Iteration 8: log likelihood = -288.21086  
Iteration 9: log likelihood = -288.21084

Multinomial logistic regression		Number of obs	=	331
		LR chi2(32)	=	73.16
		Prob > chi2	=	0.0000
Log likelihood = -288.21084		Pseudo R2	=	0.1126

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.000726	.0267104	0.03	0.978	.9497204 1.054471
exper		.9989167	.0408713	-0.03	0.979	.9219382 1.082323
pyperstudlep		.9117177	.0726315	-1.16	0.246	.7799196 1.065788
pyperstude~s		.9914182	.0162174	-0.53	0.598	.9601368 1.023719
pyperstudw~e		1.104647	.254605	0.43	0.666	.7031276 1.735454
pyperstudb~k		1.128932	.2607304	0.53	0.600	.7179265 1.775233
pyperstudl~o		1.159803	.2844999	0.60	0.546	.7171074 1.87579
pyperstudm~e		.9762203	.0305078	-0.77	0.441	.9182205 1.037884
male		3.695617	1.596782	3.03	0.002	1.584545 8.619248
age		.9848878	.0375825	-0.40	0.690	.9139145 1.061373
Asian		5.59e+08	7.04e+12	0.00	0.999	0 .
Black		8.094041	6.461743	2.62	0.009	1.692867 38.69974
Latino		6.43e-06	.004445	-0.02	0.986	0 .
Other		1.66e-06	.0021943	-0.01	0.992	0 .
AsianMale		(omitted)				
BlackMale		.2320917	.2731158	-1.24	0.215	.0231207 2.329796
LatinoMale		(omitted)				
OtherMale		6.97e+11	1.05e+15	0.02	0.986	0 .
3						
tenure		.9747997	.0174146	-1.43	0.153	.9412583 1.009536
exper		1.035522	.0293091	1.23	0.217	.9796412 1.09459
pyperstudlep		1.07782	.0588463	1.37	0.170	.96844 1.199554
pyperstude~s		1.029798	.0107573	2.81	0.005	1.008929 1.051099
pyperstudw~e		.9637217	.1665869	-0.21	0.831	.6867754 1.352348
pyperstudb~k		.9492665	.1634474	-0.30	0.762	.6773698 1.330303
pyperstudl~o		.8886496	.1602342	-0.65	0.513	.6240921 1.265355
pyperstudm~e		1.00557	.013675	0.41	0.683	.9791219 1.032733
male		3.027489	.8604349	3.90	0.000	1.734462 5.284457
age		1.010341	.0268838	0.39	0.699	.959 1.064431
Asian		.7503808	14613.62	-0.00	1.000	0 .
Black		2.577861	1.811986	1.35	0.178	.6500506 10.22284
Latino		.9863455	.9357131	-0.01	0.988	.1536465 6.331922
Other		1.95e-06	.0015755	-0.02	0.987	0 .

AsianMale		(omitted)					
BlackMale		.7290358	.7589204	-0.30	0.761	.0947665	5.608449
LatinoMale		(omitted)					
OtherMale		3.89e+11	4.25e+14	0.02	0.980	0	.

-----> region = 9

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -229.9819  
Iteration 1: log likelihood = -203.4783  
Iteration 2: log likelihood = -200.87946  
Iteration 3: log likelihood = -200.57549  
Iteration 4: log likelihood = -200.52313  
Iteration 5: log likelihood = -200.51127  
Iteration 6: log likelihood = -200.50845  
Iteration 7: log likelihood = -200.50784  
Iteration 8: log likelihood = -200.50772  
Iteration 9: log likelihood = -200.50769  
Iteration 10: log likelihood = -200.50768

Multinomial logistic regression	Number of obs	=	224
	LR chi2(30)	=	58.95
	Prob > chi2	=	0.0012
Log likelihood = -200.50768	Pseudo R2	=	0.1282

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.078408	.0404286	2.01	0.044	1.002011 1.160631
exper		.9209255	.049971	-1.52	0.129	.8280123 1.024265
pyperstudlep		.9341712	.073786	-0.86	0.389	.8001915 1.090584
pyperstude~s		.9808496	.0162617	-1.17	0.243	.9494895 1.013245
pyperstudw~e		.748235	.1387318	-1.56	0.118	.5202535 1.076121
pyperstudb~k		.7446789	.1537362	-1.43	0.153	.4968686 1.116083
pyperstudl~o		.7678157	.1401644	-1.45	0.148	.5368717 1.098104
pyperstudm~e		1.010144	.0259292	0.39	0.694	.9605806 1.062264
male		1.002576	.4540766	0.01	0.995	.4126666 2.435762
age		.9530854	.0407272	-1.12	0.261	.8765128 1.036347
Asian		(omitted)				
Black		1.05e+07	1.64e+10	0.01	0.992	0 .
Latino		2.66e+07	7.44e+10	0.01	0.995	0 .
Other		.7717484	.9981067	-0.20	0.841	.0611803 9.735083
AsianMale		(omitted)				
BlackMale		7.63e-15	1.85e-11	-0.01	0.989	0 .
LatinoMale		7.38e-08	.0005707	-0.00	0.998	0 .
OtherMale		(omitted)				
3						
tenure		.9938109	.0203453	-0.30	0.762	.9547243 1.034498
exper		1.044224	.037253	1.21	0.225	.973704 1.119852
pyperstudlep		.9979924	.0503782	-0.04	0.968	.9039803 1.101781
pyperstude~s		1.028297	.0135793	2.11	0.035	1.002023 1.055259
pyperstudw~e		1.069611	.1517546	0.47	0.635	.8099514 1.412515
pyperstudb~k		1.072768	.1681925	0.45	0.654	.7889531 1.458681
pyperstudl~o		1.030553	.1462806	0.21	0.832	.7802746 1.361109
pyperstudm~e		.9722325	.028289	-0.97	0.333	.9183383 1.029289
male		2.082202	.6934814	2.20	0.028	1.083998 3.999606
age		.9891413	.0313708	-0.34	0.731	.9295276 1.052578
Asian		(omitted)				
Black		9427637	1.47e+10	0.01	0.992	0 .

Latino		1.58e+07	4.41e+10	0.01	0.995	0	.
Other		1.76e-07	.0003188	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		2.87e-08	.0000448	-0.01	0.991	0	.
LatinoMale		.9402079	4447.313	-0.00	1.000	0	.
OtherMale		(omitted)					

-----> region = 10

Iteration 0: log likelihood = -3030.4898  
Iteration 1: log likelihood = -2872.9791  
Iteration 2: log likelihood = -2870.2981  
Iteration 3: log likelihood = -2870.2619  
Iteration 4: log likelihood = -2870.2618

Multinomial logistic regression	Number of obs	=	2861
	LR chi2(36)	=	320.46
	Prob > chi2	=	0.0000
Log likelihood = -2870.2618	Pseudo R2	=	0.0529

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.016172	.007376	2.21	0.027	1.001818 1.030732
exper		.9953053	.0096752	-0.48	0.628	.9765219 1.01445
pyperstudlep		1.00013	.0060734	0.02	0.983	.9882969 1.012105
pyperstude~s		.997733	.0036249	-0.62	0.532	.9906536 1.004863
pyperstudw~e		.9876913	.0104565	-1.17	0.242	.967408 1.0084
pyperstudb~k		.9861925	.0107449	-1.28	0.202	.9653563 1.007479
pyperstudl~o		.9887066	.0106581	-1.05	0.292	.9680361 1.009818
pyperstudm~e		.9963875	.0040335	-0.89	0.371	.9885132 1.004325
male		1.515919	.1763875	3.58	0.000	1.206793 1.904229
age		.9744642	.0080051	-3.15	0.002	.9589 .990281
Asian		2.514527	2.08841	1.11	0.267	.4937428 12.80595
Black		2.089611	.3460349	4.45	0.000	1.510457 2.890829
Latino		2.436669	.5730184	3.79	0.000	1.536823 3.863394
Other		1.52243	.7864397	0.81	0.416	.5531381 4.190264
AsianMale		.4799211	.5046773	-0.70	0.485	.0611027 3.769464
BlackMale		1.049877	.2603801	0.20	0.844	.6457012 1.707045
LatinoMale		1.295405	.4544481	0.74	0.461	.6513148 2.576441
OtherMale		.5350906	.4048544	-0.83	0.409	.1214508 2.357514
3						
tenure		1.020572	.0067825	3.06	0.002	1.007365 1.033953
exper		1.056293	.0109972	5.26	0.000	1.034958 1.078069
pyperstudlep		1.013702	.0066064	2.09	0.037	1.000836 1.026733
pyperstude~s		1.018028	.0036591	4.97	0.000	1.010882 1.025225
pyperstudw~e		1.019038	.0112572	1.71	0.088	.9972109 1.041342
pyperstudb~k		.9976924	.0115196	-0.20	0.841	.9753678 1.020528
pyperstudl~o		.9858115	.0113054	-1.25	0.213	.9639005 1.00822
pyperstudm~e		1.001679	.004209	0.40	0.690	.9934639 1.009963
male		1.28769	.147697	2.20	0.027	1.028441 1.612292
age		.9820205	.0090851	-1.96	0.050	.9643746 .9999893
Asian		1.352453	1.149209	0.36	0.722	.255763 7.151658
Black		1.34126	.2404969	1.64	0.102	.9438148 1.906071
Latino		2.2661	.53724	3.45	0.001	1.423897 3.606445
Other		1.159297	.1688898	-1.73	0.083	.019941 1.27253
AsianMale		.9984216	1.037802	-0.00	0.999	.1301784 7.657534
BlackMale		1.444586	.3896635	1.36	0.173	.851412 2.451021
LatinoMale		1.094279	.4222324	0.23	0.815	.5136743 2.331137
OtherMale		4.201751	5.274984	1.14	0.253	.3587652 49.20965

-----> region = 11

Iteration 0: log likelihood = -2175.4825  
 Iteration 1: log likelihood = -2021.1053  
 Iteration 2: log likelihood = -2018.4596  
 Iteration 3: log likelihood = -2018.4024  
 Iteration 4: log likelihood = -2018.3925  
 Iteration 5: log likelihood = -2018.3903  
 Iteration 6: log likelihood = -2018.3897  
 Iteration 7: log likelihood = -2018.3896  
 Iteration 8: log likelihood = -2018.3896

Multinomial logistic regression	Number of obs	=	2028
	LR chi2(36)	=	314.19
	Prob > chi2	=	0.0000
Log likelihood = -2018.3896	Pseudo R2	=	0.0722

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9678414	.0091242	-3.47	0.001	.9501224	.9858908
exper	.95341	.0089921	-5.06	0.000	.9359477	.9711981
pyperstudlep	1.00673	.0098907	0.68	0.495	.9875304	1.026304
pyperstude~s	1.000048	.0042381	0.01	0.991	.9917756	1.008389
pyperstudw~e	.9690651	.0165579	-1.84	0.066	.9371496	1.002068
pyperstudb~k	.9627179	.0173243	-2.11	0.035	.9293548	.9972788
pyperstudl~o	.9584966	.0168232	-2.42	0.016	.9260844	.9920431
pyperstudm~e	.9928344	.0047416	-1.51	0.132	.9835844	1.002171
male	1.092648	.1391478	0.70	0.487	.8512963	1.402427
age	1.027344	.0082365	3.36	0.001	1.011327	1.043615
Asian	2.905801	2.601308	1.19	0.233	.5026435	16.79854
Black	1.690716	.4338048	2.05	0.041	1.022513	2.795583
Latino	2.587394	.8635936	2.85	0.004	1.345109	4.977002
Other	1.057656	.7676311	0.08	0.938	.2550089	4.386659
AsianMale	.2052114	.3454937	-0.94	0.347	.0075704	5.562677
BlackMale	2.185621	.8677282	1.97	0.049	1.003768	4.759005
LatinoMale	.9795781	.5002993	-0.04	0.968	.3600024	2.665463
OtherMale	.4172221	.5670959	-0.64	0.520	.0290669	5.988745
3						
tenure	.9833049	.0080328	-2.06	0.039	.9676862	.9991757
exper	.9809384	.0090271	-2.09	0.036	.9634043	.9987916
pyperstudlep	1.003716	.0099199	0.38	0.707	.9844609	1.023348
pyperstude~s	1.028635	.0042378	6.85	0.000	1.020362	1.036974
pyperstudw~e	1.033852	.0188514	1.83	0.068	.9975565	1.071468
pyperstudb~k	1.005656	.0193716	0.29	0.770	.9683959	1.044349
pyperstudl~o	1.000178	.0189845	0.01	0.993	.9636522	1.038087
pyperstudm~e	.9995028	.0044476	-0.11	0.911	.9908236	1.008258
male	1.325745	.1650107	2.27	0.023	1.038757	1.692021
age	1.065963	.009173	7.42	0.000	1.048135	1.084094
Asian	1.097888	1.363583	0.08	0.940	.0962412	12.52435
Black	2.001456	.5563172	2.50	0.013	1.160779	3.450983
Latino	3.233939	1.122508	3.38	0.001	1.637864	6.385369
Other	.4857926	.533604	-0.66	0.511	.0564254	4.182415
AsianMale	3.25e-06	.0025089	-0.02	0.987	0	.
BlackMale	.8571814	.3988945	-0.33	0.741	.344319	2.133951
LatinoMale	.3884876	.2292525	-1.60	0.109	.1221997	1.235049
OtherMale	1.068967	1.514077	0.05	0.962	.0665777	17.16326

-----> region = 12

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -804.03991  
Iteration 1: log likelihood = -737.13954  
Iteration 2: log likelihood = -735.0526  
Iteration 3: log likelihood = -734.93129  
Iteration 4: log likelihood = -734.91068  
Iteration 5: log likelihood = -734.90859  
Iteration 6: log likelihood = -734.90814  
Iteration 7: log likelihood = -734.90804  
Iteration 8: log likelihood = -734.90802

Multinomial logistic regression                      Number of obs        =        755  
   LR chi2(32)        =        138.26  
   Prob > chi2        =        0.0000  
Log likelihood = -734.90802                      Pseudo R2        =        0.0860

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.006486	.0150003	0.43	0.664	.9775117	1.03632
	exper	.965081	.0193861	-1.77	0.077	.9278233	1.003835
pyperstudlep		.9904162	.0208185	-0.46	0.647	.9504418	1.032072
pyperstude~s		1.005441	.0078175	0.70	0.485	.990235	1.02088
pyperstudw~e		.7771932	.0460016	-4.26	0.000	.692065	.8727928
pyperstudb~k		.7772947	.0497524	-3.94	0.000	.6856504	.8811881
pyperstudl~o		.7831095	.0478653	-4.00	0.000	.6946967	.8827743
pyperstudm~e		.9890293	.0092151	-1.18	0.236	.971132	1.007256
	male	1.286724	.2835393	1.14	0.253	.8354426	1.981773
	age	.9730337	.0172232	-1.54	0.122	.9398558	1.007383
	Asian	(omitted)					
	Black	.9257301	.370853	-0.19	0.847	.4221706	2.029929
	Latino	2.965248	2.295881	1.40	0.160	.6501408	13.5243
	Other	1.76e-06	.0015859	-0.01	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	4.804162	3.116644	2.42	0.016	1.347134	17.13266
	LatinoMale	.3292934	.3324623	-1.10	0.271	.0455187	2.382187
	OtherMale	829969.5	7.49e+08	0.02	0.988	0	.
3							
	tenure	1.006517	.0117559	0.56	0.578	.9837376	1.029824
	exper	1.062592	.0203138	3.18	0.001	1.023514	1.103161
pyperstudlep		1.049895	.0204175	2.50	0.012	1.01063	1.090685
pyperstude~s		1.028669	.0070204	4.14	0.000	1.015001	1.042521
pyperstudw~e		.9071381	.0540463	-1.64	0.102	.8071602	1.0195
pyperstudb~k		.8840786	.0559872	-1.95	0.052	.7808826	1.000912
pyperstudl~o		.8600867	.0524566	-2.47	0.013	.763181	.9692971
pyperstudm~e		.9953421	.0089699	-0.52	0.604	.9779158	1.013079
	male	1.571992	.3070474	2.32	0.021	1.071989	2.305208
	age	.9589754	.0166541	-2.41	0.016	.9268833	.9921788
	Asian	(omitted)					
	Black	1.250083	.4851778	0.58	0.565	.5842161	2.674879
	Latino	.9893465	.9706139	-0.01	0.991	.1446318	6.767574
	Other	1.981938	2.61902	0.52	0.605	.1486876	26.41833
	AsianMale	(omitted)					
	BlackMale	1.398686	.9473007	0.50	0.620	.3708682	5.274984
	LatinoMale	.6393878	.8233643	-0.35	0.728	.0512436	7.977915
	OtherMale	7.25e-07	.0007456	-0.01	0.989	0	.

-----> region = 13

```

Iteration 0: log likelihood = -1337.1624
Iteration 1: log likelihood = -1279.8342
Iteration 2: log likelihood = -1278.5053
Iteration 3: log likelihood = -1278.2692
Iteration 4: log likelihood = -1278.2327
Iteration 5: log likelihood = -1278.224
Iteration 6: log likelihood = -1278.2221
Iteration 7: log likelihood = -1278.2217
Iteration 8: log likelihood = -1278.2216
Iteration 9: log likelihood = -1278.2216

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```

Multinomial logistic regression      Number of obs   =      1238
                                     LR chi2(36)        =      117.88
                                     Prob > chi2         =      0.0000
Log likelihood = -1278.2216         Pseudo R2        =      0.0441

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.023455	.0108681	2.18	0.029	1.002374	1.044979
	exper	.9880103	.0140016	-0.85	0.395	.9609453	1.015838
pyperstudlep		.9970519	.0130075	-0.23	0.821	.9718807	1.022875
pyperstude~s		.9923456	.0062496	-1.22	0.222	.9801718	1.004671
pyperstudw~e		.9657674	.0278623	-1.21	0.227	.9126734	1.02195
pyperstudb~k		.9913809	.0312852	-0.27	0.784	.9319209	1.054635
pyperstudl~o		.9774176	.027882	-0.80	0.423	.9242694	1.033622
pyperstudm~e		.989711	.006621	-1.55	0.122	.9768188	1.002773
	male	.9833002	.164808	-0.10	0.920	.7079765	1.365694
	age	.9819979	.0125008	-1.43	0.154	.9577999	1.006807
	Asian	1.411987	5046.538	0.00	1.000	0	.
	Black	1.345207	.4733999	0.84	0.399	.6748975	2.68127
	Latino	1.344248	.3640218	1.09	0.275	.7906318	2.285517
	Other	4.02e-07	.0004298	-0.01	0.989	0	.
	AsianMale	6599791	2.81e+10	0.00	0.997	0	.
	BlackMale	2.164373	1.203495	1.39	0.165	.7278187	6.436368
	LatinoMale	1.701185	.699151	1.29	0.196	.7601963	3.806952
	OtherMale	4307758	4.61e+09	0.01	0.989	0	.
3							
	tenure	.9971598	.0096408	-0.29	0.769	.9784421	1.016236
	exper	1.048405	.0147071	3.37	0.001	1.019972	1.07763
pyperstudlep		1.009361	.0130205	0.72	0.470	.9841614	1.035206
pyperstude~s		1.010031	.0061151	1.65	0.099	.9981167	1.022088
pyperstudw~e		1.046475	.0329117	1.44	0.149	.9839172	1.113011
pyperstudb~k		1.054127	.036233	1.53	0.125	.9854506	1.127589
pyperstudl~o		1.032985	.0324573	1.03	0.302	.9712888	1.0986
pyperstudm~e		.9974007	.0061306	-0.42	0.672	.9854569	1.009489
	male	1.047169	.163845	0.29	0.768	.770608	1.422983
	age	.9895184	.0129418	-0.81	0.420	.9644753	1.015212
	Asian	7985906	1.80e+10	0.01	0.994	0	.
	Black	.5379277	.2286686	-1.46	0.145	.2338219	1.23755
	Latino	.9205245	.2661456	-0.29	0.775	.5223152	1.622326
	Other	2.35e-07	.0002734	-0.01	0.990	0	.
	AsianMale	2.09e-07	.0008947	-0.00	0.997	0	.
	BlackMale	1.918002	1.322598	0.94	0.345	.4964613	7.409908
	LatinoMale	1.240311	.5612872	0.48	0.634	.5108933	3.011142
	OtherMale	6.036927	8523.883	0.00	0.999	0	.

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-----> region = 14

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note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity

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note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -295.68874  
 Iteration 1: log likelihood = -254.91488  
 Iteration 2: log likelihood = -250.32871  
 Iteration 3: log likelihood = -249.8637  
 Iteration 4: log likelihood = -249.79137  
 Iteration 5: log likelihood = -249.774  
 Iteration 6: log likelihood = -249.77037  
 Iteration 7: log likelihood = -249.76961  
 Iteration 8: log likelihood = -249.76943  
 Iteration 9: log likelihood = -249.76939  
 Iteration 10: log likelihood = -249.76938

Multinomial logistic regression

Number of obs = 312  
 LR chi2(30) = 91.84  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1553

Log likelihood = -249.76938

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9864428	.0341467	-0.39	0.693	.9217364	1.055692
exper	1.015838	.0521151	0.31	0.759	.9186614	1.123293
pyperstudlep	.8225653	.1191969	-1.35	0.178	.6191899	1.09274
pyperstude~s	.9927623	.0212722	-0.34	0.735	.951933	1.035343
pyperstudw~e	.7057239	.1566025	-1.57	0.116	.4568255	1.090233
pyperstudb~k	.7427083	.1754309	-1.26	0.208	.4674776	1.179983
pyperstudl~o	.7090606	.1565059	-1.56	0.119	.4600485	1.092856
pyperstudm~e	.985552	.0285736	-0.50	0.616	.9311103	1.043177
male	3.0486	1.477638	2.30	0.021	1.179039	7.882657
age	.9636348	.0426146	-0.84	0.402	.8836291	1.050884
Asian	(omitted)					
Black	5.266246	6.691907	1.31	0.191	.43638	63.55319
Latino	2.871969	3.682681	0.82	0.411	.2326469	35.45376
Other	1.54e-06	.0038707	-0.01	0.996	0	.
AsianMale	(omitted)					
BlackMale	1047682	1.51e+09	0.01	0.992	0	.
LatinoMale	.8365477	1.343169	-0.11	0.911	.0359578	19.46202
OtherMale	(omitted)					
3						
tenure	.9855578	.0199403	-0.72	0.472	.9472403	1.025425
exper	1.082111	.034485	2.48	0.013	1.016589	1.151856
pyperstudlep	1.066987	.0492606	1.40	0.160	.9746771	1.168039
pyperstude~s	1.04091	.0132798	3.14	0.002	1.015205	1.067266
pyperstudw~e	1.207544	.2421355	0.94	0.347	.8151189	1.788896
pyperstudb~k	1.126535	.2441637	0.55	0.583	.7366415	1.722792
pyperstudl~o	1.16647	.2314559	0.78	0.438	.7906325	1.720967
pyperstudm~e	1.012864	.0172597	0.75	0.453	.979594	1.047263
male	2.775335	.8106544	3.49	0.000	1.565618	4.919772
age	.9467833	.0273809	-1.89	0.059	.8946104	1.001999
Asian	(omitted)					
Black	.9200642	1.104787	-0.07	0.945	.0874418	9.680929
Latino	4.21e-07	.0003753	-0.02	0.987	0	.
Other	.3323227	.4899104	-0.75	0.455	.0184807	5.975882
AsianMale	(omitted)					
BlackMale	4679455	6.75e+09	0.01	0.992	0	.
LatinoMale	856817.9	7.63e+08	0.02	0.988	0	.
OtherMale	(omitted)					

-----> region = 15

note: Asian omitted because of collinearity  
 note: Black omitted because of collinearity  
 note: Other omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: BlackMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -308.17265  
 Iteration 1: log likelihood = -277.59764  
 Iteration 2: log likelihood = -276.07817  
 Iteration 3: log likelihood = -276.06641  
 Iteration 4: log likelihood = -276.06641

Multinomial logistic regression	Number of obs	=	316
	LR chi2 (24)	=	64.21
	Prob > chi2	=	0.0000
Log likelihood = -276.06641	Pseudo R2	=	0.1042

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.024154	.0302374	0.81	0.419	.9665716	1.085166
	exper	1.068476	.0543861	1.30	0.193	.967026	1.180569
pyperstudlep		.9901203	.0244263	-0.40	0.687	.9433845	1.039171
pyperstude~s		.9801283	.0143017	-1.38	0.169	.9524945	1.008564
pyperstudw~e		.6791227	.1579476	-1.66	0.096	.430506	1.071315
pyperstudb~k		.7179708	.1752348	-1.36	0.175	.4449943	1.158402
pyperstudl~o		.6994355	.160551	-1.56	0.119	.4460255	1.096821
pyperstudm~e		1.000761	.0259666	0.03	0.977	.9511399	1.052971
	male	2.059496	.9486202	1.57	0.117	.8350083	5.079619
	age	.9448705	.0443808	-1.21	0.227	.8617696	1.035985
	Asian	(omitted)					
	Black	(omitted)					
	Latino	2.105168	1.36119	1.15	0.250	.592795	7.475996
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.5364082	.4453842	-0.75	0.453	.1053745	2.730582
	OtherMale	(omitted)					
3							
	tenure	.9960303	.0193386	-0.20	0.838	.9588395	1.034664
	exper	1.074832	.0317871	2.44	0.015	1.014302	1.138975
pyperstudlep		1.016171	.0204413	0.80	0.425	.9768865	1.057036
pyperstude~s		1.028679	.0098793	2.94	0.003	1.009498	1.048226
pyperstudw~e		.9666263	.1739118	-0.19	0.850	.6793813	1.37532
pyperstudb~k		.9195317	.1728744	-0.45	0.655	.6361192	1.329214
pyperstudl~o		.9379945	.1677199	-0.36	0.720	.6606928	1.331683
pyperstudm~e		1.007734	.0188365	0.41	0.680	.971483	1.045337
	male	1.807545	.526125	2.03	0.042	1.021712	3.197788
	age	.9708264	.0261496	-1.10	0.272	.9209036	1.023456
	Asian	(omitted)					
	Black	(omitted)					
	Latino	.5308546	.3451629	-0.97	0.330	.1484302	1.89858
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.820083	1.439919	0.76	0.449	.3860839	8.580264
	OtherMale	(omitted)					

-----> region = 16

note: AsianMale omitted because of collinearity

```

Iteration 0:  log likelihood = -444.9293
Iteration 1:  log likelihood = -397.56401
Iteration 2:  log likelihood = -392.51143
Iteration 3:  log likelihood = -392.01285
Iteration 4:  log likelihood = -391.91689
Iteration 5:  log likelihood = -391.8942
Iteration 6:  log likelihood = -391.88915
Iteration 7:  log likelihood = -391.88804
Iteration 8:  log likelihood = -391.88785
Iteration 9:  log likelihood = -391.88783

```

```

Multinomial logistic regression      Number of obs   =      432
                                      LR chi2(34)        =     106.08
                                      Prob > chi2        =      0.0000
Log likelihood = -391.88783          Pseudo R2       =      0.1192

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2	tenure	1.054393	.0293942	1.90	0.057	.9983271	1.113608
	exper	.9287933	.031828	-2.16	0.031	.8684603	.9933178
	pyperstudlep	.9936927	.0299864	-0.21	0.834	.9366247	1.054238
	pyperstude~s	.9715395	.010654	-2.63	0.008	.9508809	.9926469
	pyperstudw~e	.9295636	.0374829	-1.81	0.070	.8589265	1.00601
	pyperstudb~k	.9526308	.0456584	-1.01	0.311	.8672166	1.046458
	pyperstudl~o	.9503419	.03609	-1.34	0.180	.8821751	1.023776
	pyperstudm~e	1.003744	.0122138	0.31	0.759	.9800883	1.02797
	male	1.95148	.6505351	2.01	0.045	1.015342	3.750733
	age	.9897306	.0260866	-0.39	0.695	.9398999	1.042203
	Asian	4.41e+07	2.88e+11	0.00	0.998	0	.
	Black	3.503424	3.256215	1.35	0.177	.5667009	21.65866
	Latino	3.044703	1.914329	1.77	0.077	.8878916	10.44071
	Other	6.88e-07	.0013756	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.2580683	.4024221	-0.87	0.385	.0121452	5.483608
	LatinoMale	.5097036	.5442977	-0.63	0.528	.0628557	4.133245
	OtherMale	3034968	6.07e+09	0.01	0.994	0	.
3	tenure	1.018975	.0170037	1.13	0.260	.986187	1.052852
	exper	1.00704	.023967	0.29	0.768	.9611443	1.055127
	pyperstudlep	1.070883	.0224908	3.26	0.001	1.027697	1.115884
	pyperstude~s	1.002026	.0074944	0.27	0.787	.9874442	1.016823
	pyperstudw~e	1.046706	.0408455	1.17	0.242	.9696353	1.129903
	pyperstudb~k	1.031439	.0477636	0.67	0.504	.9419467	1.129434
	pyperstudl~o	1.027388	.0387803	0.72	0.474	.9541237	1.106278
	pyperstudm~e	.9987041	.0100759	-0.13	0.898	.9791497	1.018649
	male	1.426215	.3419855	1.48	0.139	.8914136	2.281869
	age	1.034085	.0222902	1.55	0.120	.991307	1.078709
	Asian	1.836161	17925.68	0.00	1.000	0	.
	Black	6.22e-07	.0005908	-0.02	0.988	0	.
	Latino	.0518031	.0743441	-2.06	0.039	.0031101	.8628578
	Other	4.32e-07	.0006614	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	1.242469	1787.937	0.00	1.000	0	.
	LatinoMale	8.093602	14.95205	1.13	0.258	.2165904	302.4436
	OtherMale	.5834715	1079.941	-0.00	1.000	0	.

```
-----> region = 17
```

```

Iteration 0:  log likelihood = -467.25763
Iteration 1:  log likelihood = -412.76624

```

```

Iteration 2: log likelihood = -409.50171
Iteration 3: log likelihood = -408.93812
Iteration 4: log likelihood = -408.8484
Iteration 5: log likelihood = -408.83905
Iteration 6: log likelihood = -408.83714
Iteration 7: log likelihood = -408.83674
Iteration 8: log likelihood = -408.83665
Iteration 9: log likelihood = -408.83662

```

```

Multinomial logistic regression
Log likelihood = -408.83662
Number of obs   =      469
LR chi2(36)    =     116.84
Prob > chi2    =      0.0000
Pseudo R2     =      0.1250

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9916835	.0244067	-0.34	0.734	.9449827	1.040692
	exper	.9799141	.0325487	-0.61	0.541	.9181521	1.045831
pyperstudlep		1.066942	.0395701	1.75	0.081	.9921375	1.147386
pyperstude~s		.9790255	.0139304	-1.49	0.136	.9520995	1.006713
pyperstudw~e		.7991742	.1344676	-1.33	0.183	.5746718	1.111381
pyperstudb~k		.8391267	.1398785	-1.05	0.293	.605252	1.163373
pyperstudl~o		.7976603	.1331018	-1.35	0.175	.5751511	1.106252
pyperstudm~e		.9817851	.013549	-1.33	0.183	.9555855	1.008703
	male	2.999037	1.078153	3.06	0.002	1.482433	6.067204
	age	1.009643	.0286476	0.34	0.735	.9550277	1.067382
	Asian	6.26e-07	.0025737	-0.00	0.997	0	.
	Black	8.80e-07	.0008768	-0.01	0.989	0	.
	Latino	1.02e-06	.000666	-0.02	0.983	0	.
	Other	2.11929	7455.007	0.00	1.000	0	.
	AsianMale	1.53e+07	9.86e+10	0.00	0.998	0	.
	BlackMale	1.03e+12	1.46e+15	0.02	0.984	0	.
	LatinoMale	2787586	1.83e+09	0.02	0.982	0	.
	OtherMale	2.67518	9410.456	0.00	1.000	0	.
3							
	tenure	.9779084	.0154107	-1.42	0.156	.9481658	1.008584
	exper	1.053838	.0250137	2.21	0.027	1.005935	1.104023
pyperstudlep		1.061482	.0247048	2.56	0.010	1.014149	1.111024
pyperstude~s		1.01765	.0107408	1.66	0.097	.9968142	1.03892
pyperstudw~e		1.172331	.1688858	1.10	0.270	.8839476	1.554799
pyperstudb~k		1.155186	.166658	1.00	0.317	.8706619	1.53269
pyperstudl~o		1.150641	.1647348	0.98	0.327	.8691103	1.523367
pyperstudm~e		.9831185	.0096374	-1.74	0.082	.9644099	1.00219
	male	2.604849	.6427722	3.88	0.000	1.60598	4.224983
	age	1.015843	.0217475	0.73	0.463	.9741008	1.059375
	Asian	6.18e-07	.0014602	-0.01	0.995	0	.
	Black	3.005947	2.934524	1.13	0.260	.4436134	20.36845
	Latino	1.277976	.7233352	0.43	0.665	.4214464	3.875282
	Other	5513678	9.20e+09	0.01	0.993	0	.
	AsianMale	1.96e+13	6.59e+16	0.01	0.993	0	.
	BlackMale	1364608	1.37e+09	0.01	0.989	0	.
	LatinoMale	2.109677	1.688817	0.93	0.351	.4393604	10.13004
	OtherMale	1.18e-06	.0019635	-0.01	0.993	0	.

```
-----> region = 18
```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -442.5883
Iteration 1: log likelihood = -391.42695

```

```

Iteration 2: log likelihood = -389.45785
Iteration 3: log likelihood = -389.24676
Iteration 4: log likelihood = -389.21299
Iteration 5: log likelihood = -389.20535
Iteration 6: log likelihood = -389.20347
Iteration 7: log likelihood = -389.20311
Iteration 8: log likelihood = -389.20305
Iteration 9: log likelihood = -389.20303

```

Multinomial logistic regression

```

Number of obs   =      432
LR chi2(32)     =      106.77
Prob > chi2     =      0.0000
Pseudo R2      =      0.1206

```

Log likelihood = -389.20303

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.987593	.023191	-0.53	0.595	.9431696	1.034109
	exper	.9517814	.0325068	-1.45	0.148	.8901549	1.017674
pyperstudlep		.9735488	.0175191	-1.49	0.136	.9398104	1.008498
pyperstude~s		1.052091	.0146261	3.65	0.000	1.023811	1.081151
pyperstudw~e		.748157	.1115034	-1.95	0.052	.5586397	1.001968
pyperstudb~k		.7524119	.1152373	-1.86	0.063	.5572987	1.015835
pyperstudl~o		.7311514	.1095342	-2.09	0.037	.5451152	.9806779
pyperstudm~e		.9928173	.0118324	-0.60	0.545	.969895	1.016281
	male	1.946242	.6666739	1.94	0.052	.9945407	3.808649
	age	.983353	.0289958	-0.57	0.569	.9281333	1.041858
	Asian	(omitted)					
	Black	2.589632	3.818473	0.65	0.519	.1439201	46.59665
	Latino	1.752383	.9818237	1.00	0.317	.5844111	5.254599
	Other	3.258534	9356.162	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	1.66e-07	.0001602	-0.02	0.987	0	.
	LatinoMale	.8740364	.6125436	-0.19	0.848	.2213039	3.451994
	OtherMale	5.08e-08	.0001719	-0.00	0.996	0	.
3							
	tenure	.9812405	.0156658	-1.19	0.236	.9510115	1.01243
	exper	1.098533	.033621	3.07	0.002	1.034575	1.166446
pyperstudlep		.9787935	.0143718	-1.46	0.144	.9510268	1.007371
pyperstude~s		1.05747	.0118986	4.97	0.000	1.034405	1.08105
pyperstudw~e		.9311026	.1439111	-0.46	0.644	.6877585	1.260547
pyperstudb~k		.9336376	.1471296	-0.44	0.663	.6855507	1.271502
pyperstudl~o		.9133077	.140821	-0.59	0.556	.6751063	1.235555
pyperstudm~e		.9859984	.0102108	-1.36	0.173	.9661874	1.006216
	male	1.689703	.4668597	1.90	0.058	.9831644	2.903987
	age	.9619793	.026638	-1.40	0.162	.9111612	1.015632
	Asian	(omitted)					
	Black	3.369968	4.032991	1.02	0.310	.3228158	35.18007
	Latino	1.845633	.8653674	1.31	0.191	.736275	4.62648
	Other	2.97e+07	4.21e+10	0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	.2897751	.4720712	-0.76	0.447	.0118952	7.0591
	LatinoMale	.302612	.1917814	-1.89	0.059	.0873845	1.047944
	OtherMale	7.06e-14	1.40e-10	-0.02	0.988	0	.

-----> region = 19

```

Iteration 0: log likelihood = -782.6971
Iteration 1: log likelihood = -724.91861
Iteration 2: log likelihood = -722.95618
Iteration 3: log likelihood = -722.89148

```

```

Iteration 4: log likelihood = -722.88028
Iteration 5: log likelihood = -722.87777
Iteration 6: log likelihood = -722.87716
Iteration 7: log likelihood = -722.87703
Iteration 8: log likelihood = -722.87701

```

```

Multinomial logistic regression
Number of obs   =      732
LR chi2(36)     =     119.64
Prob > chi2     =      0.0000
Pseudo R2      =      0.0764

Log likelihood = -722.87701

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.01312	.0136808	0.97	0.334	.9866577	1.040292
exper	1.056491	.0199631	2.91	0.004	1.01808	1.096351
pypersistudlep	1.011567	.0063895	1.82	0.069	.9991215	1.024168
pypersistude~s	1.003453	.0076258	0.45	0.650	.9886174	1.018511
pypersistudw~e	.9384748	.1102007	-0.54	0.589	.7455384	1.181341
pypersistudb~k	.9346891	.1179895	-0.54	0.593	.7298214	1.197065
pypersistudl~o	.9312359	.1058789	-0.63	0.531	.7452131	1.163694
pypersistudm~e	1.002708	.0075719	0.36	0.720	.9879769	1.017659
male	1.242983	.3272662	0.83	0.409	.7419107	2.082471
age	.9705472	.0153582	-1.89	0.059	.9409077	1.00112
Asian	1079724	9.22e+08	0.02	0.987	0	.
Black	2.237943	1.78286	1.01	0.312	.4696111	10.66497
Latino	1.358353	.3416863	1.22	0.223	.8296553	2.223964
Other	3.09e-06	.0023692	-0.02	0.987	0	.
AsianMale	4.22e-07	.0009525	-0.01	0.995	0	.
BlackMale	1.257277	1.412573	0.20	0.839	.1390244	11.37027
LatinoMale	.8843603	.317367	-0.34	0.732	.4376852	1.786885
OtherMale	880577.4	1.96e+09	0.01	0.995	0	.
3						
tenure	1.03237	.015308	2.15	0.032	1.002799	1.062813
exper	1.115367	.0265442	4.59	0.000	1.064536	1.168626
pypersistudlep	1.020724	.0074343	2.82	0.005	1.006256	1.035399
pypersistude~s	1.015121	.0096122	1.58	0.113	.9964547	1.034136
pypersistudw~e	.9608759	.1183893	-0.32	0.746	.7547288	1.22333
pypersistudb~k	.936779	.1236077	-0.49	0.621	.7233049	1.213257
pypersistudl~o	.9296748	.1105631	-0.61	0.540	.7363775	1.173712
pypersistudm~e	1.009734	.0083635	1.17	0.242	.9934741	1.02626
male	.8269646	.2659361	-0.59	0.555	.4403081	1.553164
age	.9683606	.0195857	-1.59	0.112	.9307243	1.007519
Asian	.8629643	1240.429	-0.00	1.000	0	.
Black	1.233948	1.191828	0.22	0.828	.1858449	8.193006
Latino	1.23549	.3585604	0.73	0.466	.699529	2.182089
Other	4.69e-06	.0042538	-0.01	0.989	0	.
AsianMale	1258727	2.50e+09	0.01	0.994	0	.
BlackMale	3.904427	5.121031	1.04	0.299	.2986239	51.04931
LatinoMale	1.34837	.5775524	0.70	0.485	.5823858	3.121817
OtherMale	6.50e+12	1.07e+16	0.02	0.986	0	.

```

-----> region = 20

```

```

Iteration 0: log likelihood = -2070.0149
Iteration 1: log likelihood = -1956.8899
Iteration 2: log likelihood = -1950.9192
Iteration 3: log likelihood = -1950.504
Iteration 4: log likelihood = -1950.4396
Iteration 5: log likelihood = -1950.4242
Iteration 6: log likelihood = -1950.4208

```

```
Iteration 7: log likelihood = -1950.4202
Iteration 8: log likelihood = -1950.42
Iteration 9: log likelihood = -1950.42
Iteration 10: log likelihood = -1950.42
```

```
Multinomial logistic regression      Number of obs   =      2115
                                     LR chi2(36)        =      239.19
                                     Prob > chi2         =      0.0000
Log likelihood = -1950.42            Pseudo R2        =      0.0578
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9972861	.0085225	-0.32	0.750	.9807214	1.014131
	exper	1.017316	.0116382	1.50	0.133	.9947598	1.040385
pyperstudlep		.9790828	.0075108	-2.76	0.006	.9644721	.9939149
pyperstude~s		1.004104	.0052541	0.78	0.434	.9938586	1.014455
pyperstudw~e		.9678276	.0433093	-0.73	0.465	.886559	1.056546
pyperstudb~k		.9666313	.045785	-0.72	0.474	.8809337	1.060666
pyperstudl~o		.9672319	.0436514	-0.74	0.460	.8853514	1.056685
pyperstudm~e		.9920198	.0049589	-1.60	0.109	.9823481	1.001787
	male	1.542916	.2260739	2.96	0.003	1.157766	2.056191
	age	.9843281	.0096833	-1.61	0.108	.965531	1.003491
	Asian	2.04e-07	.0003553	-0.01	0.993	0	.
	Black	1.594196	.4943772	1.50	0.133	.8681102	2.927577
	Latino	1.558628	.2600873	2.66	0.008	1.123836	2.161633
	Other	.7565837	.8368368	-0.25	0.801	.0865697	6.612232
	AsianMale	.9857568	1.72e+10	0.01	0.993	0	.
	BlackMale	.9125124	.4139309	-0.20	0.840	.3750762	2.220026
	LatinoMale	1.027714	.2496324	0.11	0.910	.6384328	1.654359
	OtherMale	3.68e-07	.0006081	-0.01	0.993	0	.
3							
	tenure	1.028129	.0090168	3.16	0.002	1.010608	1.045955
	exper	1.087522	.015814	5.77	0.000	1.056964	1.118963
pyperstudlep		1.01544	.006871	2.26	0.024	1.002062	1.028997
pyperstude~s		1.030719	.0059605	5.23	0.000	1.019103	1.042468
pyperstudw~e		1.152646	.0626128	2.62	0.009	1.036234	1.282136
pyperstudb~k		1.119771	.0635472	1.99	0.046	1.001898	1.251512
pyperstudl~o		1.11391	.0607499	1.98	0.048	1.000985	1.239574
pyperstudm~e		1.007838	.0043229	1.82	0.069	.9994009	1.016347
	male	1.652238	.2615271	3.17	0.002	1.211543	2.253234
	age	.9586188	.0123885	-3.27	0.001	.9346428	.98321
	Asian	2.08e-07	.0003684	-0.01	0.993	0	.
	Black	1.172461	.3806911	0.49	0.624	.6204662	2.215534
	Latino	1.020579	.1847253	0.11	0.910	.7157787	1.455173
	Other	4.90e-07	.0006628	-0.01	0.991	0	.
	AsianMale	1.565448	5133.104	0.00	1.000	0	.
	BlackMale	.9352446	.4598901	-0.14	0.892	.3567482	2.45182
	LatinoMale	1.078473	.282123	0.29	0.773	.6458615	1.800857
	OtherMale	2961538	4.00e+09	0.01	0.991	0	.

```
by pyschooltype:
. bysort pyschooltype: mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age A
> sian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert1999a==1,
rrr
```

```
-----> pyschooltype =
```

```
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -458.75142
```

```

Iteration 1: log likelihood = -403.09744
Iteration 2: log likelihood = -399.75966
Iteration 3: log likelihood = -399.28653
Iteration 4: log likelihood = -399.21006
Iteration 5: log likelihood = -399.19245
Iteration 6: log likelihood = -399.18813
Iteration 7: log likelihood = -399.18729
Iteration 8: log likelihood = -399.18715
Iteration 9: log likelihood = -399.18712
Iteration 10: log likelihood = -399.18711

```

```

Multinomial logistic regression      Number of obs   =      441
                                     LR chi2(32)        =    119.13
                                     Prob > chi2         =    0.0000
Log likelihood = -399.18711          Pseudo R2        =    0.1298

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.007995	.0203609	0.39	0.693	.9688679	1.048702
	exper	1.017965	.024646	0.74	0.462	.9707875	1.067434
pyperstudlep		1.001795	.0121032	0.15	0.882	.9783519	1.0258
pyperstude~s		.9967214	.0081838	-0.40	0.689	.9808099	1.012891
pyperstudw~e		.914562	.0391361	-2.09	0.037	.8409853	.9945758
pyperstudb~k		.9149733	.0410309	-1.98	0.048	.8379871	.9990323
pyperstudl~o		.9134234	.0394117	-2.10	0.036	.839354	.9940291
pyperstudm~e		.9976486	.0099126	-0.24	0.813	.9784083	1.017267
	male	1.750914	.5311359	1.85	0.065	.9661724	3.173035
	age	.9668222	.0195684	-1.67	0.096	.9292196	1.005946
	Asian	1.20e-07	.0006026	-0.00	0.997	0	.
	Black	1.21596	.6075933	0.39	0.696	.4566561	3.237795
	Latino	.5342821	.280987	-1.19	0.233	.1905948	1.497719
	Other	1.14e+07	1.95e+10	0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	1.08e+07	1.05e+10	0.02	0.987	0	.
	LatinoMale	1.115748	.8342598	0.15	0.884	.257698	4.830822
	OtherMale	(omitted)					
3							
	tenure	1.012668	.0177211	0.72	0.472	.9785244	1.048003
	exper	1.136532	.0333916	4.36	0.000	1.072934	1.203899
pyperstudlep		1.019142	.0111843	1.73	0.084	.9974556	1.041301
pyperstude~s		1.016888	.0091351	1.86	0.062	.9991406	1.034951
pyperstudw~e		1.029279	.05322	0.56	0.577	.9300806	1.139056
pyperstudb~k		.9931466	.0541905	-0.13	0.900	.8924173	1.105245
pyperstudl~o		1.000704	.0526961	0.01	0.989	.9025725	1.109504
pyperstudm~e		.9977827	.0125119	-0.18	0.859	.9735587	1.022609
	male	1.576706	.4990032	1.44	0.150	.8479274	2.931858
	age	.9562291	.02466	-1.74	0.083	.9090976	1.005804
	Asian	6.32e-08	.000306	-0.00	0.997	0	.
	Black	1.031915	.56844	0.06	0.955	.3505514	3.03764
	Latino	.5765967	.316867	-1.00	0.316	.1963798	1.692963
	Other	.9146457	1.57e+10	0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	1.52e+07	1.47e+10	0.02	0.986	0	.
	LatinoMale	.9609075	.705041	-0.05	0.957	.2281048	4.04789
	OtherMale	(omitted)					

```

-----> pyschooltype = B

```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -673.84496

```



```

Iteration 1: log likelihood = -637.60658
Iteration 2: log likelihood = -633.37957
Iteration 3: log likelihood = -633.29905
Iteration 4: log likelihood = -633.28059
Iteration 5: log likelihood = -633.27659
Iteration 6: log likelihood = -633.27576
Iteration 7: log likelihood = -633.27558
Iteration 8: log likelihood = -633.27553
Iteration 9: log likelihood = -633.27553

```

```

Multinomial logistic regression      Number of obs   =      699
                                     LR chi2(34)       =      81.14
                                     Prob > chi2      =      0.0000
Log likelihood = -633.27553         Pseudo R2       =      0.0602

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.01903	.0198147	0.97	0.332	.9809247	1.058616
	exper	.9684606	.0227859	-1.36	0.173	.9248151	1.014166
pyperstudlep		1.009737	.0107895	0.91	0.365	.9888095	1.031107
pyperstude~s		.9901459	.007119	-1.38	0.168	.9762908	1.004198
pyperstudw~e		.9435889	.0361797	-1.51	0.130	.8752769	1.017232
pyperstudb~k		.9592357	.0377559	-1.06	0.290	.8880178	1.036165
pyperstudl~o		.9504053	.0367739	-1.31	0.189	.880995	1.025284
pyperstudm~e		.9961848	.0042912	-0.89	0.375	.9878097	1.004631
	male	1.719261	.5398077	1.73	0.084	.9291472	3.18126
	age	.9713542	.0193946	-1.46	0.145	.9340756	1.010121
	Asian	2.261536	6700.218	0.00	1.000	0	.
	Black	3.096384	1.479417	2.37	0.018	1.213837	7.898583
	Latino	6.668828	3.821797	3.31	0.001	2.168901	20.50498
	Other	7.45e-06	.0065592	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	.4782581	.3313133	-1.06	0.287	.1230247	1.859226
	LatinoMale	.2800446	.1967582	-1.81	0.070	.0706605	1.109885
	OtherMale	149011.3	1.31e+08	0.01	0.989	0	.
3							
	tenure	.998807	.0117503	-0.10	0.919	.9760404	1.022105
	exper	1.045561	.0183184	2.54	0.011	1.010267	1.082088
pyperstudlep		1.003845	.009657	0.40	0.690	.9850954	1.022952
pyperstude~s		1.011197	.0053039	2.12	0.034	1.000854	1.021646
pyperstudw~e		1.063679	.040896	1.61	0.108	.9864699	1.146931
pyperstudb~k		1.05631	.0416351	1.39	0.165	.9777792	1.141148
pyperstudl~o		1.050287	.0407425	1.26	0.206	.9733929	1.133254
pyperstudm~e		1.002417	.0032514	0.74	0.457	.9960645	1.00881
	male	1.088082	.2099433	0.44	0.662	.7454593	1.588179
	age	.9618811	.0155017	-2.41	0.016	.9319731	.9927489
	Asian	2599785	3.41e+09	0.01	0.991	0	.
	Black	.5799293	.2827805	-1.12	0.264	.2230073	1.508104
	Latino	1.28206	.6968558	0.46	0.648	.4418174	3.720264
	Other	1.100972	1.374499	0.08	0.939	.0953019	12.71895
	AsianMale	(omitted)					
	BlackMale	2.103928	1.373754	1.14	0.255	.5851147	7.565204
	LatinoMale	.824893	.5363251	-0.30	0.767	.2306574	2.95004
	OtherMale	.4174628	.7048255	-0.52	0.605	.0152576	11.42216

```

-----> pyschooltype = E

```

```

Iteration 0: log likelihood = -9397.1584
Iteration 1: log likelihood = -8821.3799
Iteration 2: log likelihood = -8815.2444

```

Iteration 3: log likelihood = -8815.2405  
 Iteration 4: log likelihood = -8815.2405

Multinomial logistic regression	Number of obs	=	8842
	LR chi2(36)	=	1163.84
	Prob > chi2	=	0.0000
Log likelihood = -8815.2405	Pseudo R2	=	0.0619

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9952514	.0047113	-1.01	0.315	.9860602	1.004528
	exper	.9978462	.0058668	-0.37	0.714	.9864135	1.009411
pyperstudlep		1.011664	.0022209	5.28	0.000	1.00732	1.016026
pyperstude~s		.9861744	.002226	-6.17	0.000	.9818213	.9905469
pyperstudw~e		.9802937	.0070861	-2.75	0.006	.9665032	.994281
pyperstudb~k		.9814006	.0075627	-2.44	0.015	.9666893	.9963358
pyperstudl~o		.9840268	.007253	-2.18	0.029	.9699134	.9983456
pyperstudm~e		1.013233	.0040741	3.27	0.001	1.005279	1.021249
	male	1.790914	.1637323	6.37	0.000	1.497113	2.142372
	age	.990068	.0047602	-2.08	0.038	.980782	.9994418
	Asian	1.422356	.662323	0.76	0.449	.5710108	3.543007
	Black	1.516765	.1666092	3.79	0.000	1.222976	1.881129
	Latino	1.138033	.1098411	1.34	0.180	.9418859	1.375028
	Other	.8476186	.3567933	-0.39	0.695	.3714489	1.934203
	AsianMale	.6036273	.5036299	-0.61	0.545	.1176463	3.097131
	BlackMale	.8586897	.1852134	-0.71	0.480	.5626496	1.310492
	LatinoMale	.6878945	.1169625	-2.20	0.028	.492938	.9599563
	OtherMale	.1945041	.2238977	-1.42	0.155	.0203748	1.856798
3							
	tenure	1.006863	.003795	1.81	0.070	.9994529	1.014329
	exper	1.058321	.0056926	10.54	0.000	1.047223	1.069538
pyperstudlep		1.00431	.0019775	2.18	0.029	1.000442	1.008194
pyperstude~s		1.005379	.0020277	2.66	0.008	1.001413	1.009361
pyperstudw~e		1.010915	.0074253	1.48	0.139	.996466	1.025574
pyperstudb~k		.9986235	.0077876	-0.18	0.860	.9834762	1.014004
pyperstudl~o		.9988614	.007516	-0.15	0.880	.9842383	1.013702
pyperstudm~e		1.001972	.0036852	0.54	0.592	.994775	1.009221
	male	2.931429	.2223135	14.18	0.000	2.526539	3.401204
	age	1.004073	.0046611	0.88	0.381	.9949793	1.013251
	Asian	.5212498	.3154665	-1.08	0.282	.1591801	1.706881
	Black	1.036744	.1101717	0.34	0.734	.8418152	1.276809
	Latino	1.258867	.1104984	2.62	0.009	1.059899	1.495185
	Other	.4643445	.2083009	-1.71	0.087	.1927519	1.118618
	AsianMale	.9254521	.8403266	-0.09	0.932	.1561163	5.48605
	BlackMale	.6165526	.1248209	-2.39	0.017	.4146147	.9168443
	LatinoMale	.5126161	.0759501	-4.51	0.000	.3834217	.6853426
	OtherMale	.7541338	.5566445	-0.38	0.702	.1774826	3.204358

-----> pyschooltype = M

Iteration 0: log likelihood = -5339.7108  
 Iteration 1: log likelihood = -5076.8337  
 Iteration 2: log likelihood = -5071.2774  
 Iteration 3: log likelihood = -5070.9542  
 Iteration 4: log likelihood = -5070.8839  
 Iteration 5: log likelihood = -5070.8688  
 Iteration 6: log likelihood = -5070.8662  
 Iteration 7: log likelihood = -5070.8659  
 Iteration 8: log likelihood = -5070.8659  
 Iteration 9: log likelihood = -5070.8659

Multinomial logistic regression

Number of obs = 5096  
 LR chi2(36) = 537.69  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0503

Log likelihood = -5070.8659

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.009603	.0054376	1.77	0.076	.9990019	1.020317
	exper	.9954023	.0066435	-0.69	0.490	.982466	1.008509
	pyperstudlep	1.013285	.0038587	3.47	0.001	1.00575	1.020876
	pyperstude~s	.9955557	.0028425	-1.56	0.119	.9900002	1.001142
	pyperstudw~e	.976599	.0090367	-2.56	0.010	.959047	.9944723
	pyperstudb~k	.9793154	.0098465	-2.08	0.038	.9602055	.9988056
	pyperstudl~o	.9750126	.0094248	-2.62	0.009	.9567141	.9936611
	pyperstudm~e	.9987118	.0038114	-0.34	0.736	.9912694	1.00621
	male	1.503531	.1247813	4.91	0.000	1.277819	1.769113
	age	.9758725	.0054112	-4.40	0.000	.9653241	.9865362
	Asian	.3138238	.3388901	-1.07	0.283	.0378003	2.60541
	Black	1.207991	.1653839	1.38	0.168	.9236922	1.579793
	Latino	1.863592	.251622	4.61	0.000	1.430282	2.428175
	Other	.5059864	.3283698	-1.05	0.294	.1418191	1.805273
	AsianMale	2.24e-06	.0017163	-0.02	0.986	0	.
	BlackMale	1.72686	.3376926	2.79	0.005	1.17707	2.533449
	LatinoMale	1.019033	.1775923	0.11	0.914	.7241774	1.433942
	OtherMale	2.542942	2.184666	1.09	0.277	.4721292	13.69657
3							
	tenure	.9996765	.0053086	-0.06	0.951	.9893259	1.010135
	exper	1.060236	.0083412	7.43	0.000	1.044013	1.076711
	pyperstudlep	1.002939	.0046865	0.63	0.530	.9937957	1.012167
	pyperstude~s	1.005352	.0032824	1.63	0.102	.9989391	1.011806
	pyperstudw~e	1.023991	.0125763	1.93	0.054	.9996361	1.048939
	pyperstudb~k	1.009767	.0134339	0.73	0.465	.9837778	1.036444
	pyperstudl~o	1.010421	.0129382	0.81	0.418	.9853783	1.036101
	pyperstudm~e	1.000574	.0042768	0.13	0.893	.9922265	1.008992
	male	2.350314	.2118637	9.48	0.000	1.969683	2.8045
	age	.9757434	.0069684	-3.44	0.001	.9621806	.9894973
	Asian	8.22e-07	.0006879	-0.02	0.987	0	.
	Black	.9551083	.1803397	-0.24	0.808	.6596783	1.382844
	Latino	1.886346	.3114747	3.84	0.000	1.364805	2.607188
	Other	.3607752	.3776089	-0.97	0.330	.0463787	2.806435
	AsianMale	2342079	1.96e+09	0.02	0.986	0	.
	BlackMale	1.657877	.4096342	2.05	0.041	1.021491	2.690731
	LatinoMale	.8135895	.1650879	-1.02	0.309	.5466199	1.210947
	OtherMale	3.656306	4.358852	1.09	0.277	.3534155	37.82678

-----> pyschooltype = S

Iteration 0: log likelihood = -6312.0399  
 Iteration 1: log likelihood = -6027.5559  
 Iteration 2: log likelihood = -6014.203  
 Iteration 3: log likelihood = -6014.1143  
 Iteration 4: log likelihood = -6014.1143

Multinomial logistic regression

Number of obs = 6635  
 LR chi2(36) = 595.85  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0472

Log likelihood = -6014.1143

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9993064	.0041924	-0.17	0.869	.9911231	1.007557
exper	.993769	.0055211	-1.13	0.261	.9830066	1.004649
pyperstudlep	1.01219	.0040325	3.04	0.002	1.004317	1.020124
pyperstude~s	.9948341	.0024036	-2.14	0.032	.9901343	.9995562
pyperstudw~e	.9821815	.0072932	-2.42	0.015	.9679905	.9965805
pyperstudb~k	.9854158	.0078177	-1.85	0.064	.9702119	1.000858
pyperstudl~o	.9833626	.0075741	-2.18	0.029	.9686292	.9983202
pyperstudm~e	.9987261	.0024381	-0.52	0.602	.993959	1.003516
male	1.252153	.0906368	3.11	0.002	1.086534	1.443017
age	.9879724	.0049928	-2.39	0.017	.978235	.9978067
Asian	1.891347	1.390303	0.87	0.386	.4477804	7.988725
Black	1.708391	.2369562	3.86	0.000	1.301741	2.242075
Latino	1.620486	.2162345	3.62	0.000	1.247563	2.104883
Other	.8257573	.4719922	-0.33	0.738	.2693498	2.53156
AsianMale	.995853	.9116857	-0.00	0.996	.1655523	5.990391
BlackMale	1.443698	.2578109	2.06	0.040	1.017353	2.048714
LatinoMale	1.414658	.2267336	2.16	0.030	1.033298	1.936767
OtherMale	1.305197	.9144694	0.38	0.704	.3305923	5.152992
3						
tenure	.9787778	.0046393	-4.53	0.000	.969727	.9879131
exper	1.054393	.0077885	7.17	0.000	1.039238	1.069769
pyperstudlep	.9975754	.0056777	-0.43	0.670	.9865091	1.008766
pyperstude~s	1.014115	.0029116	4.88	0.000	1.008424	1.019837
pyperstudw~e	1.057497	.0143836	4.11	0.000	1.029678	1.086067
pyperstudb~k	1.040408	.0148542	2.77	0.006	1.011698	1.069933
pyperstudl~o	1.040521	.0144711	2.86	0.004	1.012541	1.069274
pyperstudm~e	1.018086	.0023221	7.86	0.000	1.013545	1.022647
male	2.083776	.1945763	7.86	0.000	1.735276	2.502266
age	.9881835	.0070927	-1.66	0.098	.9743794	1.002183
Asian	1.453945	1.604856	0.34	0.735	.1671078	12.65026
Black	1.167752	.2680765	0.68	0.499	.744635	1.831293
Latino	1.238277	.252484	1.05	0.295	.8303442	1.846619
Other	1.703024	1.102583	0.82	0.411	.4787735	6.05775
AsianMale	.6843264	.934644	-0.28	0.781	.0470656	9.95
BlackMale	1.010782	.2875538	0.04	0.970	.578765	1.765277
LatinoMale	1.134106	.2647435	0.54	0.590	.7177132	1.792076
OtherMale	.5659241	.4557779	-0.71	0.480	.1167418	2.743404

### 1999-00

```
. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino Oth
> er AsianMale BlackMale LatinoMale OtherMale if validcert2000a==1, rrr
```

```
Iteration 0: log likelihood = -23057.491
Iteration 1: log likelihood = -21669.451
Iteration 2: log likelihood = -21596
Iteration 3: log likelihood = -21595.461
Iteration 4: log likelihood = -21595.461
```

```
Multinomial logistic regression      Number of obs   =      21876
LR chi2(36)                        =      2924.06
Prob > chi2                         =      0.0000
Pseudo R2                          =      0.0634
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				

2							
	tenure	1.003834	.0025718	1.49	0.135	.9988057	1.008887
	exper	.9907986	.0031742	-2.89	0.004	.9845968	.9970395
pyperstudlep		1.012367	.0013557	9.18	0.000	1.009713	1.015027
pyperstude~s		.993964	.001022	-5.89	0.000	.991963	.9959691
pyperstudw~e		1.00011	.0000451	2.43	0.015	1.000021	1.000198
pyperstudb~k		1.000294	.0000737	3.99	0.000	1.00015	1.000439
pyperstudl~o		1.000119	.0000413	2.87	0.004	1.000038	1.0002
pyperstudm~e		1.000096	.0014843	0.06	0.948	.9971914	1.00301
	male	1.270898	.053829	5.66	0.000	1.169655	1.380904
	age	.9801086	.0027145	-7.25	0.000	.9748028	.9854434
	Asian	.9877393	.3155904	-0.04	0.969	.5280521	1.8476
	Black	1.350515	.0909794	4.46	0.000	1.183469	1.541139
	Latino	1.30275	.0822214	4.19	0.000	1.151168	1.474292
	Other	.66472	.1911854	-1.42	0.156	.3782839	1.168045
	AsianMale	1.099701	.5383855	0.19	0.846	.4212577	2.870788
	BlackMale	1.409787	.1463307	3.31	0.001	1.150276	1.727846
	LatinoMale	1.39004	.1235262	3.71	0.000	1.167845	1.65451
	OtherMale	1.866377	.7331536	1.59	0.112	.8642294	4.030602

3								
	tenure		1.010633	.0023872	4.48	0.000	1.005965	1.015323
	exper		1.037545	.0034985	10.93	0.000	1.03071	1.044424
pyperstudlep		1.010852	.0013836	7.89	0.000	1.008144	1.013568	
pyperstude~s		.9999961	.0010425	-0.00	0.997	.9979549	1.002041	
pyperstudw~e		.9990893	.0000672	-13.54	0.000	.9989576	.9992211	
pyperstudb~k		.9986852	.0001105	-11.89	0.000	.9984687	.9989018	
pyperstudl~o		.9986377	.0000656	-20.75	0.000	.9985092	.9987663	
pyperstudm~e		.9938899	.0013154	-4.63	0.000	.9913151	.9964714	
	male		1.580474	.0650831	11.12	0.000	1.457926	1.713324
	age		.9933403	.0030494	-2.18	0.030	.9873815	.9993351
	Asian		.4845283	.2247892	-1.56	0.118	.1951721	1.202875
	Black		.8970605	.0688678	-1.42	0.157	.7717464	1.042723
	Latino		1.166772	.0765022	2.35	0.019	1.026065	1.326774
	Other		.4167353	.1475466	-2.47	0.013	.2082042	.8341252
	AsianMale		1.691863	1.050639	0.85	0.397	.5009251	5.714228
	BlackMale		1.157958	.1393116	1.22	0.223	.9147162	1.465882
	LatinoMale		1.058005	.1000871	0.60	0.551	.8789505	1.273536
	OtherMale		2.199655	.9964048	1.74	0.082	.9052643	5.344829

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=21876)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
	tenure	20.127	2	0.000
	exper	166.782	2	0.000
	pyperstudlep	108.726	2	0.000
	pyperstude~s	38.438	2	0.000
	pyperstudw~e	256.657	2	0.000
	pyperstudb~k	224.319	2	0.000
	pyperstudl~o	671.761	2	0.000
	pyperstudm~e	25.221	2	0.000
	male	129.164	2	0.000
	age	53.505	2	0.000
	Asian	2.938	2	0.230
	Black	27.466	2	0.000
	Latino	18.578	2	0.000
	Other	7.961	2	0.019
	AsianMale	0.732	2	0.694
	BlackMale	10.982	2	0.004
	LatinoMale	14.505	2	0.001
	OtherMale	4.576	2	0.101

-----  
 \*\*\*\* Wald tests for independent variables (N=21876)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
-----+-----			
tenure	20.052	2	0.000
exper	160.822	2	0.000
pyperstudlep	106.554	2	0.000
pyperstude~s	38.399	2	0.000
pyperstudw~e	215.919	2	0.000
pyperstudb~k	190.729	2	0.000
pyperstudl~o	491.412	2	0.000
pyperstudm~e	24.588	2	0.000
male	128.675	2	0.000
age	52.630	2	0.000
Asian	2.521	2	0.284
Black	27.933	2	0.000
Latino	18.643	2	0.000
Other	7.138	2	0.028
AsianMale	0.718	2	0.699
BlackMale	10.952	2	0.004
LatinoMale	14.494	2	0.001
OtherMale	4.454	2	0.108
-----			

\*\*\*\* Hausman tests of IIA assumption (N=21876)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
-----+-----				
2	-96.390	17	---	---
3	-18.651	17	---	---
-----				

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=21876)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
-----+-----						
2	-4894.266	-4881.732	25.068	19	0.158	for Ho
3	-5052.604	-5044.125	16.958	19	0.593	for Ho
-----						

\*\*\*\* Wald tests for combining alternatives (N=21876)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
-----+-----			
2- 3	1885.958	18	0.000
2- 1	574.030	18	0.000
3- 1	1480.595	18	0.000
-----			

\*\*\*\* LR tests for combining alternatives (N=21876)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	2318.572	18	0.000
2-	1	593.155	18	0.000
3-	1	1841.249	18	0.000

by region

```
. bysort region: mlogit admin tenure exper pyperstudlep pyperstuddecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert2000a==1, rrr
```

-----> region = 1

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -1595.9549
Iteration 1: log likelihood = -1444.4133
Iteration 2: log likelihood = -1434.7892
Iteration 3: log likelihood = -1434.4243
Iteration 4: log likelihood = -1434.3438
Iteration 5: log likelihood = -1434.3263
Iteration 6: log likelihood = -1434.3228
Iteration 7: log likelihood = -1434.322
Iteration 8: log likelihood = -1434.3218
Iteration 9: log likelihood = -1434.3217
Iteration 10: log likelihood = -1434.3217
```

Multinomial logistic regression

```
Number of obs   =      1494
LR chi2(34)     =      323.27
Prob > chi2     =      0.0000
Pseudo R2      =      0.1013
```

Log likelihood = -1434.3217

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	tenure	1.010008	.0099184	1.01	0.311	.990754 1.029636
	exper	1.034054	.0140244	2.47	0.014	1.006929 1.06191
	pyperstudlep	1.003124	.0036016	0.87	0.385	.9960896 1.010208
	pyperstude~s	1.005231	.0065435	0.80	0.423	.992488 1.018139
	pyperstudw~e	1.001418	.0013442	1.06	0.291	.9987871 1.004056
	pyperstudb~k	.9911399	.0099259	-0.89	0.374	.9718752 1.010786
	pyperstudl~o	1.000087	.0001091	0.80	0.426	.9998731 1.000301
	pyperstudm~e	.9877618	.0076269	-1.59	0.111	.9729258 1.002824
	male	1.481651	.4155749	1.40	0.161	.8550672 2.56739
	age	.9575338	.0113298	-3.67	0.000	.9355832 .9799993
	Asian	.676388	.799612	-0.33	0.741	.0666688 6.862287
	Black	4.14e+07	1.65e+11	0.00	0.996	0 .
	Latino	1.231526	.2717799	0.94	0.345	.7990897 1.897979
	Other	.9877328	1.22999	-0.01	0.992	.0860334 11.33996
	AsianMale	(omitted)				
	BlackMale	4.34e-08	.000173	-0.00	0.997	0 .
	LatinoMale	1.386759	.4349199	1.04	0.297	.7499705 2.564233
	OtherMale	4475370	5.30e+09	0.01	0.990	0 .
3						
	tenure	1.035472	.0106284	3.40	0.001	1.014849 1.056515
	exper	1.094473	.0188393	5.24	0.000	1.058165 1.132028
	pyperstudlep	1.011727	.0037828	3.12	0.002	1.004339 1.019168
	pyperstude~s	.9985759	.007879	-0.18	0.857	.983252 1.014139
	pyperstudw~e	.9993466	.0019681	-0.33	0.740	.9954967 1.003211
	pyperstudb~k	.9898933	.0141083	-0.71	0.476	.9626241 1.017935
	pyperstudl~o	.9988035	.0001637	-7.31	0.000	.9984827 .9991244
	pyperstudm~e	.9989393	.005682	-0.19	0.852	.9878646 1.010138
	male	.5811001	.2195661	-1.44	0.151	.2770946 1.218635

age		.9503866	.0149373	-3.24	0.001	.9215564	.9801187
Asian		3.04e-07	.0003788	-0.01	0.990	0	.
Black		3.566104	23645.68	0.00	1.000	0	.
Latino		1.164131	.2670431	0.66	0.508	.7425784	1.824993
Other		8.61e-07	.0012856	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		1.24e-07	.0008557	-0.00	0.998	0	.
LatinoMale		3.299506	1.350342	2.92	0.004	1.479414	7.358821
OtherMale		1.55e+13	2.96e+16	0.02	0.987	0	.

-----> region = 2

Iteration 0: log likelihood = -827.15811  
Iteration 1: log likelihood = -773.30022  
Iteration 2: log likelihood = -770.13063  
Iteration 3: log likelihood = -769.94982  
Iteration 4: log likelihood = -769.91244  
Iteration 5: log likelihood = -769.9041  
Iteration 6: log likelihood = -769.90273  
Iteration 7: log likelihood = -769.90258  
Iteration 8: log likelihood = -769.90255  
Iteration 9: log likelihood = -769.90255

Multinomial logistic regression	Number of obs	=	860
	LR chi2(36)	=	114.51
	Prob > chi2	=	0.0000
Log likelihood = -769.90255	Pseudo R2	=	0.0692

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.011835	.0157494	0.76	0.450	.9814327 1.043179
exper		.95339	.0179529	-2.53	0.011	.9188443 .9892345
pypersistudlep		1.002526	.0146815	0.17	0.863	.9741595 1.031718
pypersistude~s		1.006977	.0071968	0.97	0.331	.9929695 1.021181
pypersistudw~e		1.0001	.0005185	0.19	0.846	.9990847 1.001117
pypersistudb~k		1.001135	.0028213	0.40	0.687	.9956207 1.00668
pypersistudl~o		.9999984	.0003027	-0.01	0.996	.9994053 1.000592
pypersistudm~e		.984547	.0082706	-1.85	0.064	.9684696 1.000891
male		1.462754	.403504	1.38	0.168	.851855 2.511755
age		1.005219	.0154925	0.34	0.736	.9753083 1.036047
Asian		2.34e+08	1.84e+12	0.00	0.998	0 .
Black		2.645504	2.086556	1.23	0.217	.5638333 12.4127
Latino		.9679961	.2588201	-0.12	0.903	.5731677 1.634803
Other		3.91e-07	.0008527	-0.01	0.995	0 .
AsianMale		1.11e-15	9.31e-12	-0.00	0.997	0 .
BlackMale		.6343548	.6573783	-0.44	0.661	.0832218 4.835347
LatinoMale		1.247354	.4881116	0.56	0.572	.5792948 2.685839
OtherMale		3100628	6.76e+09	0.01	0.995	0 .
3						
tenure		1.009451	.0123297	0.77	0.441	.9855718 1.033908
exper		1.03861	.0186	2.12	0.034	1.002788 1.075713
pypersistudlep		.982922	.0129779	-1.30	0.192	.9578121 1.00869
pypersistude~s		1.022064	.006614	3.37	0.001	1.009183 1.03511
pypersistudw~e		1.000135	.0005865	0.23	0.818	.998986 1.001285
pypersistudb~k		.9998016	.0033793	-0.06	0.953	.9932003 1.006447
pypersistudl~o		.9983896	.0003865	-4.16	0.000	.9976323 .9991475
pypersistudm~e		.988124	.007001	-1.69	0.092	.9744971 1.001942
male		1.299328	.3371534	1.01	0.313	.7813518 2.160683
age		.9782435	.0161395	-1.33	0.182	.9471167 1.010393
Asian		2.65431	38655.69	0.00	1.000	0 .



Black		2.248567	1.561162	1.17	0.243	.5766648	8.767752
Latino		.7815748	.1860175	-1.04	0.300	.4902094	1.246119
Other		3.63e-07	.0006369	-0.01	0.993	0	.
AsianMale		1.48e-06	.0219159	-0.00	0.999	0	.
BlackMale		.7836047	.7717581	-0.25	0.804	.1137009	5.400451
LatinoMale		1.679044	.6066179	1.43	0.151	.8270533	3.408715
OtherMale		3.528959	7216.514	0.00	1.000	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -351.71127  
Iteration 1: log likelihood = -316.20024  
Iteration 2: log likelihood = -314.75544  
Iteration 3: log likelihood = -314.30545  
Iteration 4: log likelihood = -314.27684  
Iteration 5: log likelihood = -314.27542  
Iteration 6: log likelihood = -314.27534  
Iteration 7: log likelihood = -314.27534

Multinomial logistic regression	Number of obs	=	335
	LR chi2(30)	=	74.87
	Prob > chi2	=	0.0000
Log likelihood = -314.27534	Pseudo R2	=	0.1064

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.026476	.0232921	1.15	0.249	.9818242 1.073158
exper		.9751117	.0334527	-0.73	0.463	.9117014 1.042932
pyperstudlep		.9373402	.0479844	-1.26	0.206	.8478567 1.036268
pyperstude~s		1.00268	.0124026	0.22	0.829	.9786638 1.027286
pyperstudw~e		.9999592	.0008503	-0.05	0.962	.9982942 1.001627
pyperstudb~k		.9999618	.002873	-0.01	0.989	.9943466 1.005609
pyperstudl~o		1.000464	.0009408	0.49	0.622	.9986218 1.00231
pyperstudm~e		.9971813	.0143343	-0.20	0.844	.9694786 1.025676
male		.8630155	.3218811	-0.39	0.693	.4154739 1.792641
age		.9531214	.029639	-1.54	0.123	.8967649 1.01302
Asian		(omitted)				
Black		2.055834	1.562967	0.95	0.343	.4632912 9.12267
Latino		2.280443	1.458158	1.29	0.197	.6512461 7.98534
Other		1.71e+09	3.19e+13	0.00	0.999	0
AsianMale		(omitted)				
BlackMale		2.228554	3.380636	0.53	0.597	.1139675 43.5778
LatinoMale		.9680384	.985637	-0.03	0.975	.1315881 7.121453
OtherMale		(omitted)				
-----						
3						
tenure		1.010832	.0174967	0.62	0.534	.9771144 1.045713
exper		.996675	.0271506	-0.12	0.903	.9448564 1.051336
pyperstudlep		1.02302	.0338933	0.69	0.492	.9587008 1.091653
pyperstude~s		1.023102	.010797	2.16	0.030	1.002157 1.044484
pyperstudw~e		1.000992	.00084	1.18	0.238	.9993465 1.002639
pyperstudb~k		.9947632	.00278	-1.88	0.060	.9893294 1.000227
pyperstudl~o		.9976044	.0009628	-2.49	0.013	.995719 .9994933
pyperstudm~e		.9980183	.0103254	-0.19	0.848	.9779847 1.018462
male		2.779895	.849993	3.34	0.001	1.526726 5.06169
age		.9940591	.0260505	-0.23	0.820	.9442901 1.046451
Asian		(omitted)				
Black		2.419144	1.858318	1.15	0.250	.5367752 10.90262

Latino		2.065686	1.286102	1.17	0.244	.6096824	6.998822
Other		8.19e+08	1.53e+13	0.00	0.999	0	.
AsianMale		(omitted)					
BlackMale		1.30e-09	.0000361	-0.00	0.999	0	.
LatinoMale		1.092327	.9864028	0.10	0.922	.1860772	6.412275
OtherMale		(omitted)					

-----> region = 4

Iteration 0: log likelihood = -4582.5444  
Iteration 1: log likelihood = -4275.1992  
Iteration 2: log likelihood = -4254.9667  
Iteration 3: log likelihood = -4254.8053  
Iteration 4: log likelihood = -4254.8053

Multinomial logistic regression	Number of obs	=	4415
	LR chi2(36)	=	655.48
	Prob > chi2	=	0.0000
Log likelihood = -4254.8053	Pseudo R2	=	0.0715

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.013454	.0056007	2.42	0.016	1.002536 1.024491
exper		1.025585	.0077006	3.36	0.001	1.010602 1.040789
pyperstudlep		1.009418	.0033244	2.85	0.004	1.002923 1.015954
pyperstude~s		.9910917	.0021644	-4.10	0.000	.9868588 .9953429
pyperstudw~e		.9997995	.0000079	-2.54	0.011	.9996447 .9999543
pyperstudb~k		1.000126	.000121	1.04	0.297	.999889 1.000364
pyperstudl~o		1.00016	.0000927	1.73	0.084	.9999787 1.000342
pyperstudm~e		1.00275	.0035282	0.78	0.435	.9958587 1.009689
male		1.133381	.1070061	1.33	0.185	.941914 1.363768
age		.9548119	.0061599	-7.17	0.000	.9428147 .9669617
Asian		.6536492	.3936997	-0.71	0.480	.2007508 2.128297
Black		.8057879	.0906008	-1.92	0.055	.6464185 1.004449
Latino		1.403377	.2274116	2.09	0.037	1.021505 1.928005
Other		.4996244	.3332651	-1.04	0.298	.1351654 1.846808
AsianMale		.2770696	.3383552	-1.05	0.293	.0252993 3.034376
BlackMale		1.76188	.3076739	3.24	0.001	1.251217 2.480962
LatinoMale		1.24393	.3362731	0.81	0.419	.732301 2.113014
OtherMale		2.710805	2.79881	0.97	0.334	.3583133 20.50849
3						
tenure		1.038312	.0064916	6.01	0.000	1.025666 1.051113
exper		1.074431	.0102555	7.52	0.000	1.054518 1.094721
pyperstudlep		1.011806	.0039398	3.01	0.003	1.004114 1.019558
pyperstude~s		.9993952	.0026338	-0.23	0.818	.9942464 1.004571
pyperstudw~e		.9991508	.0001316	-6.45	0.000	.9988929 .9994089
pyperstudb~k		.9991327	.0001952	-4.44	0.000	.9987501 .9995154
pyperstudl~o		.9989392	.0001664	-6.37	0.000	.9986131 .9992654
pyperstudm~e		.9944631	.0039925	-1.38	0.167	.9866686 1.002319
male		1.353675	.1516877	2.70	0.007	1.086756 1.686153
age		.964644	.0081994	-4.23	0.000	.9487066 .9808492
Asian		.2440195	.266215	-1.29	0.196	.0287607 2.070379
Black		.4491626	.0656949	-5.47	0.000	.3372145 .5982752
Latino		1.481788	.2709539	2.15	0.032	1.035472 2.120476
Other		.7370221	.5047988	-0.45	0.656	.192521 2.821518
AsianMale		.930939	1.43135	-0.05	0.963	.0457267 18.95275
BlackMale		1.364807	.3139909	1.35	0.176	.8694442 2.142401
LatinoMale		1.335967	.4222769	0.92	0.359	.7190272 2.482255
OtherMale		1.451657	1.95498	0.28	0.782	.103641 20.33277

-----> region = 5

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -572.07006  
 Iteration 1: log likelihood = -505.64533  
 Iteration 2: log likelihood = -502.53353  
 Iteration 3: log likelihood = -502.45158  
 Iteration 4: log likelihood = -502.436  
 Iteration 5: log likelihood = -502.43229  
 Iteration 6: log likelihood = -502.43146  
 Iteration 7: log likelihood = -502.43128  
 Iteration 8: log likelihood = -502.43125

Multinomial logistic regression	Number of obs	=	572
	LR chi2(32)	=	139.28
	Prob > chi2	=	0.0000
Log likelihood = -502.43125	Pseudo R2	=	0.1217

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	1.001576	.0174699	0.09	0.928	.9679141 1.036408
	exper	1.009692	.0253421	0.38	0.701	.9612242 1.060604
	pyperstudlep	.9253927	.0412687	-1.74	0.082	.8479417 1.009918
	pyperstude~s	1.024425	.0081723	3.02	0.002	1.008532 1.040568
	pyperstudw~e	1.00108	.0004231	2.55	0.011	1.000251 1.001909
	pyperstudb~k	.999047	.0004361	-2.18	0.029	.9981927 .9999021
	pyperstudl~o	1.005149	.0030027	1.72	0.086	.9992806 1.011051
	pyperstudm~e	1.019941	.0159426	1.26	0.207	.9891674 1.051671
	male	3.053749	.8632069	3.95	0.000	1.754782 5.314268
	age	.9505744	.0200256	-2.41	0.016	.9121242 .9906455
	Asian	(omitted)				
	Black	1.528009	.5906161	1.10	0.273	.7163312 3.259402
	Latino	9.211873	13.98341	1.46	0.144	.4701582 180.4895
	Other	2.85e-06	.0030044	-0.01	0.990	0 .
	AsianMale	(omitted)				
	BlackMale	1.447616	.7546251	0.71	0.478	.521113 4.021376
	LatinoMale	.0925451	.1846498	-1.19	0.233	.0018535 4.620684
	OtherMale	.6201418	925.5313	-0.00	1.000	0 .
3						
	tenure	.9939155	.0143202	-0.42	0.672	.9662411 1.022383
	exper	1.086698	.0268142	3.37	0.001	1.035393 1.140544
	pyperstudlep	.9774465	.0396279	-0.56	0.574	.9027831 1.058285
	pyperstude~s	1.012198	.0075172	1.63	0.103	.9975712 1.027039
	pyperstudw~e	.9992186	.0004831	-1.62	0.106	.9982721 1.000166
	pyperstudb~k	.9976918	.0006203	-3.72	0.000	.9964768 .9989083
	pyperstudl~o	1.001748	.0031975	0.55	0.584	.9955007 1.008035
	pyperstudm~e	1.016748	.0149579	1.13	0.259	.9878494 1.046491
	male	3.02934	.7785837	4.31	0.000	1.830531 5.013246
	age	.969533	.0216731	-1.38	0.166	.9279717 1.012956
	Asian	(omitted)				
	Black	1.215938	.4883417	0.49	0.626	.5534191 2.671583
	Latino	4.116141	6.179791	0.94	0.346	.2170371 78.06322
	Other	.0000114	.0112624	-0.01	0.991	0 .
	AsianMale	(omitted)				
	BlackMale	.9589342	.5512492	-0.07	0.942	.3107924 2.958743
	LatinoMale	.096454	.1899497	-1.19	0.235	.0020325 4.577378
	OtherMale	.2566301	359.9483	-0.00	0.999	0 .

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -779.48985  
 Iteration 1: log likelihood = -700.65474  
 Iteration 2: log likelihood = -697.6075  
 Iteration 3: log likelihood = -697.48061  
 Iteration 4: log likelihood = -697.45324  
 Iteration 5: log likelihood = -697.44687  
 Iteration 6: log likelihood = -697.44555  
 Iteration 7: log likelihood = -697.44533  
 Iteration 8: log likelihood = -697.4453

Multinomial logistic regression

Number of obs = 730  
 LR chi2(34) = 164.09  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1053

Log likelihood = -697.4453

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9627973	.0142467	-2.56	0.010	.9352753	.9911292
	exper	.9860759	.0187867	-0.74	0.462	.9499337	1.023593
pyperstudlep		1.072216	.0204883	3.65	0.000	1.032802	1.113134
pyperstude~s		1.008806	.0077749	1.14	0.255	.9936824	1.024161
pyperstudw~e		1.000635	.0002665	2.38	0.017	1.000113	1.001157
pyperstudb~k		1.001018	.0009149	1.11	0.265	.9992269	1.002813
pyperstudl~o		.9992717	.0006715	-1.08	0.278	.9979565	1.000589
pyperstudm~e		1.001536	.0093537	0.16	0.869	.9833697	1.020038
	male	1.372676	.2889976	1.50	0.132	.9085721	2.073846
	age	.978737	.0162872	-1.29	0.197	.9473296	1.011186
	Asian	1.07e-06	.0015693	-0.01	0.993	0	.
	Black	1.045182	.5573566	0.08	0.934	.3675171	2.972391
	Latino	.1854036	.2109666	-1.48	0.139	.0199322	1.724572
	Other	.6316004	1.188546	-0.24	0.807	.0157997	25.24844
	AsianMale	(omitted)					
	BlackMale	3.207643	2.558381	1.46	0.144	.6718572	15.31422
	LatinoMale	40.08428	65.61339	2.25	0.024	1.620519	991.5027
	OtherMale	1.08e-06	.0010597	-0.01	0.989	0	.
3							
	tenure	.9643946	.0124915	-2.80	0.005	.9402199	.9891908
	exper	1.089311	.0231878	4.02	0.000	1.044799	1.13572
pyperstudlep		1.060443	.0201439	3.09	0.002	1.021687	1.100668
pyperstude~s		1.010338	.0072028	1.44	0.149	.9963185	1.024554
pyperstudw~e		.9990258	.000359	-2.71	0.007	.9983226	.9997296
pyperstudb~k		.9998707	.0010743	-0.12	0.904	.9977674	1.001978
pyperstudl~o		.9984039	.0010494	-1.52	0.129	.9963493	1.000463
pyperstudm~e		.9965542	.0074569	-0.46	0.645	.9820455	1.011277
	male	1.899949	.3894306	3.13	0.002	1.27137	2.839303
	age	.9447924	.018397	-2.92	0.004	.9094142	.9815468
	Asian	5.53e-06	.0077505	-0.01	0.993	0	.
	Black	1.242742	.6396189	0.42	0.673	.4531911	3.40785
	Latino	.6935712	.6282007	-0.40	0.686	.1175214	4.093221
	Other	5.39e-07	.0003574	-0.02	0.983	0	.
	AsianMale	(omitted)					
	BlackMale	.9554188	.7899968	-0.06	0.956	.18896	4.830784
	LatinoMale	5.006975	7.845453	1.03	0.304	.2321794	107.976
	OtherMale	337433	2.24e+08	0.02	0.985	0	.

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -973.68208  
 Iteration 1: log likelihood = -892.46174  
 Iteration 2: log likelihood = -885.65312  
 Iteration 3: log likelihood = -885.30275  
 Iteration 4: log likelihood = -885.22886  
 Iteration 5: log likelihood = -885.21072  
 Iteration 6: log likelihood = -885.20716  
 Iteration 7: log likelihood = -885.20653  
 Iteration 8: log likelihood = -885.20646  
 Iteration 9: log likelihood = -885.20645

Multinomial logistic regression	Number of obs	=	986
	LR chi2(34)	=	176.95
	Prob > chi2	=	0.0000
Log likelihood = -885.20645	Pseudo R2	=	0.0909

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9862048	.0134102	-1.02	0.307	.9602684	1.012842
	exper	1.018268	.0181471	1.02	0.310	.9833147	1.054465
pyperstudlep		1.004063	.0203255	0.20	0.841	.9650055	1.044701
pyperstude~s		1.008812	.0079083	1.12	0.263	.9934306	1.024432
pyperstudw~e		1.001176	.0003876	3.04	0.002	1.000417	1.001936
pyperstudb~k		1.000041	.0005227	0.08	0.938	.9990166	1.001066
pyperstudl~o		.9993502	.0012236	-0.53	0.595	.9969549	1.001751
pyperstudm~e		.9835354	.0163127	-1.00	0.317	.9520772	1.016033
	male	1.97902	.4388477	3.08	0.002	1.281434	3.056357
	age	.9577757	.0158252	-2.61	0.009	.9272557	.9893003
	Asian	4.43e-07	.0007666	-0.01	0.993	0	.
	Black	2.951092	1.012604	3.15	0.002	1.506299	5.781682
	Latino	1.289775	1.469008	0.22	0.823	.138365	12.0227
	Other	1.29e-06	.001867	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	1.128489	.5658312	0.24	0.809	.4223764	3.015053
	LatinoMale	5.314773	9.067022	0.98	0.327	.1876428	150.535
	OtherMale	.5262037	1012.404	-0.00	1.000	0	.
3							
	tenure	.9847027	.0098821	-1.54	0.125	.9655233	1.004263
	exper	1.053927	.0149517	3.70	0.000	1.025026	1.083643
pyperstudlep		1.010457	.0155351	0.68	0.499	.980463	1.041369
pyperstude~s		1.009031	.0057017	1.59	0.112	.9979177	1.020269
pyperstudw~e		.9988354	.0004551	-2.56	0.011	.9979439	.9997278
pyperstudb~k		.9983726	.0007307	-2.23	0.026	.9969415	.9998059
pyperstudl~o		.9977014	.0014699	-1.56	0.118	.9948246	1.000586
pyperstudm~e		1.003173	.0072574	0.44	0.661	.9890488	1.017498
	male	1.640034	.2781659	2.92	0.004	1.176198	2.286783
	age	.9852064	.0129325	-1.14	0.256	.9601825	1.010882
	Asian	7.288662	12.95455	1.12	0.264	.2237479	237.4306
	Black	1.219906	.4054233	0.60	0.550	.6359722	2.339992
	Latino	6.03e-07	.0004916	-0.02	0.986	0	.
	Other	1.537171	1.931145	0.34	0.732	.1310267	18.0337
	AsianMale	(omitted)					
	BlackMale	1.533271	.7651205	0.86	0.392	.5765792	4.077356
	LatinoMale	1.11e+07	9.07e+09	0.02	0.984	0	.
	OtherMale	.9764542	1.610766	-0.01	0.988	.0385038	24.76285

-----> region = 8

Multinomial logistic regression	Number of obs	=	327
	LR chi2(32)	=	78.61
	Prob > chi2	=	0.0000
Log likelihood = -283.02489	Pseudo R2	=	0.1219

```
-----> region = 9
```



```

Iteration 0: log likelihood = -3158.1555
Iteration 1: log likelihood = -2948.4363
Iteration 2: log likelihood = -2938.0219
Iteration 3: log likelihood = -2937.9418
Iteration 4: log likelihood = -2937.9418

```

```

Multinomial logistic regression
Number of obs   =      2969
LR chi2(36)     =      440.43
Prob > chi2     =      0.0000
Pseudo R2      =      0.0697

Log likelihood = -2937.9418

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013634	.0069201	1.98	0.047	1.000161	1.027289
	exper	.9885896	.0092071	-1.23	0.218	.9707077	1.006801
pyperstudlep		1.004739	.0044603	1.07	0.287	.9960351	1.013519
pyperstude~s		.9976149	.0029047	-0.82	0.412	.991938	1.003324
pyperstudw~e		1.000099	.0001312	0.75	0.451	.9998419	1.000356
pyperstudb~k		1.000003	.0002015	0.02	0.987	.9996084	1.000398
pyperstudl~o		.9999117	.0001929	-0.46	0.647	.9995336	1.00029
pyperstudm~e		.9962492	.0045504	-0.82	0.411	.9873704	1.005208
	male	1.481562	.1697091	3.43	0.001	1.183633	1.854483
	age	.970566	.0076611	-3.78	0.000	.9556661	.9856981
	Asian	1.341322	1.000965	0.39	0.694	.3106858	5.790881
	Black	1.953045	.3057827	4.28	0.000	1.436951	2.654498
	Latino	2.381046	.5502973	3.75	0.000	1.513706	3.745363
	Other	1.094924	.5919067	0.17	0.867	.3795175	3.158903
	AsianMale	1.968234	1.922657	0.69	0.488	.2901254	13.35265
	BlackMale	1.042855	.2483985	0.18	0.860	.6538458	1.663306
	LatinoMale	1.281687	.4340043	0.73	0.464	.6600072	2.488946
	OtherMale	.7408755	.5539981	-0.40	0.688	.1710998	3.208048
3							
	tenure	1.02139	.0066013	3.27	0.001	1.008534	1.034411
	exper	1.052014	.010879	4.90	0.000	1.030907	1.073554
pyperstudlep		.9996418	.0048673	-0.07	0.941	.9901474	1.009227
pyperstude~s		.999729	.0029682	-0.09	0.927	.9939283	1.005564
pyperstudw~e		.9989181	.0001824	-5.93	0.000	.9985606	.9992758
pyperstudb~k		.9983155	.0002758	-6.10	0.000	.997775	.9988563
pyperstudl~o		.9987175	.0002921	-4.39	0.000	.9981452	.9992901
pyperstudm~e		.9915147	.0039913	-2.12	0.034	.9837227	.9993684
	male	1.520756	.175981	3.62	0.000	1.212158	1.90792
	age	.973492	.0089159	-2.93	0.003	.956173	.9911246
	Asian	1.058851	.8210065	0.07	0.941	.2316514	4.839885
	Black	1.095774	.1929498	0.52	0.603	.7759555	1.547409
	Latino	2.49434	.5801265	3.93	0.000	1.581195	3.934827
	Other	.3336367	.2661898	-1.38	0.169	.0698471	1.593674
	AsianMale	1.495663	1.52678	0.39	0.693	.2022661	11.05972
	BlackMale	1.148366	.318667	0.50	0.618	.6666149	1.978269
	LatinoMale	.7820669	.2943695	-0.65	0.514	.3739833	1.635444
	OtherMale	1.185244	1.240549	0.16	0.871	.1523659	9.219929

```

-----> region = 11

```

```

Iteration 0: log likelihood = -2208.7232
Iteration 1: log likelihood = -1997.548
Iteration 2: log likelihood = -1989.7612
Iteration 3: log likelihood = -1989.6428
Iteration 4: log likelihood = -1989.6233
Iteration 5: log likelihood = -1989.6201

```



```

Iteration 6: log likelihood = -1989.6193
Iteration 7: log likelihood = -1989.6192
Iteration 8: log likelihood = -1989.6191
Iteration 9: log likelihood = -1989.6191

```

```

Multinomial logistic regression
Log likelihood = -1989.6191
Number of obs   =      2049
LR chi2(36)     =      438.21
Prob > chi2     =      0.0000
Pseudo R2      =      0.0992

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.96636	.0091209	-3.63	0.000	.9486478	.9844029
exper	.9521226	.0084255	-5.54	0.000	.9357513	.9687802
pyperstudlep	.9995448	.0064148	-0.07	0.943	.9870508	1.012197
pyperstude~s	.9936406	.0039271	-1.61	0.106	.9859734	1.001367
pyperstudw~e	.9999992	.0001192	-0.01	0.995	.9997656	1.000233
pyperstudb~k	.9994759	.0003038	-1.72	0.085	.9988808	1.000071
pyperstudl~o	1.000308	.0003579	0.86	0.390	.9996065	1.001009
pyperstudm~e	.9957317	.0046477	-0.92	0.359	.986664	1.004883
male	1.285061	.1652489	1.95	0.051	.9987704	1.653414
age	1.015192	.0077274	1.98	0.048	1.000159	1.030451
Asian	1.469376	1.25812	0.45	0.653	.2743534	7.869647
Black	2.144486	.5307778	3.08	0.002	1.320212	3.483394
Latino	1.905647	.6109492	2.01	0.044	1.016595	3.572211
Other	.8424714	.5961162	-0.24	0.809	.2105047	3.371698
AsianMale	.2577976	.3877458	-0.90	0.367	.0135211	4.915253
BlackMale	1.050169	.4125228	0.12	0.901	.486288	2.267903
LatinoMale	.9587277	.4709672	-0.09	0.932	.3660576	2.510968
OtherMale	.8226349	.9462061	-0.17	0.865	.0863263	7.839186
3						
tenure	.9866023	.0080726	-1.65	0.099	.9709065	1.002552
exper	.9777602	.008554	-2.57	0.010	.9611376	.9946703
pyperstudlep	.9871744	.0067929	-1.88	0.061	.9739499	1.000578
pyperstude~s	1.002461	.0037925	0.65	0.516	.9950554	1.009922
pyperstudw~e	.9986771	.0001842	-7.18	0.000	.9983161	.9990383
pyperstudb~k	.997932	.0004044	-5.11	0.000	.9971396	.998725
pyperstudl~o	.9992548	.0004641	-1.60	0.109	.9983456	1.000165
pyperstudm~e	.9874996	.0040832	-3.04	0.002	.9795291	.995535
male	1.667995	.2132501	4.00	0.000	1.298285	2.142986
age	1.059735	.008776	7.01	0.000	1.042673	1.077076
Asian	.7468349	.8747601	-0.25	0.803	.0752005	7.417
Black	1.839485	.513389	2.18	0.029	1.064465	3.178787
Latino	2.800643	.9250761	3.12	0.002	1.465881	5.350777
Other	.3102586	.3394126	-1.07	0.285	.0363528	2.64795
AsianMale	3.82e-06	.0025978	-0.02	0.985	0	.
BlackMale	.7565958	.3428547	-0.62	0.538	.3112707	1.839033
LatinoMale	.386789	.2148619	-1.71	0.087	.1302058	1.148994
OtherMale	1.709624	2.415783	0.38	0.704	.1071793	27.27032

```

-----> region = 12

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -820.52956
Iteration 1: log likelihood = -758.50063
Iteration 2: log likelihood = -755.21169
Iteration 3: log likelihood = -755.00145
Iteration 4: log likelihood = -754.96505
Iteration 5: log likelihood = -754.95698

```

```

Iteration 6: log likelihood = -754.95506
Iteration 7: log likelihood = -754.95464
Iteration 8: log likelihood = -754.95456
Iteration 9: log likelihood = -754.95454

```

```

Multinomial logistic regression
Log likelihood = -754.95454
Number of obs   =      781
LR chi2(32)     =     131.15
Prob > chi2     =     0.0000
Pseudo R2      =     0.0799

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.017023	.0148929	1.15	0.249	.9882484	1.046636
exper	.952302	.0185861	-2.50	0.012	.9165619	.9894357
pyperstudlep	1.029416	.0211455	1.41	0.158	.9887946	1.071706
pyperstude~s	1.008953	.0065757	1.37	0.171	.9961469	1.021924
pyperstudw~e	1.001013	.0004651	2.18	0.029	1.000102	1.001925
pyperstudb~k	1.001464	.0006088	2.41	0.016	1.000271	1.002658
pyperstudl~o	.9990222	.0010999	-0.89	0.374	.9968687	1.00118
pyperstudm~e	1.002464	.0082742	0.30	0.766	.9863769	1.018812
male	1.39155	.3004807	1.53	0.126	.9113743	2.124715
age	.9930421	.0167523	-0.41	0.679	.9607452	1.026425
Asian	(omitted)					
Black	.9013594	.3850072	-0.24	0.808	.3902251	2.082
Latino	2.662063	1.682911	1.55	0.121	.7710865	9.190382
Other	1.04e-06	.000913	-0.02	0.987	0	.
AsianMale	(omitted)					
BlackMale	2.703245	1.781222	1.51	0.131	.7430392	9.834655
LatinoMale	.7392438	.6850349	-0.33	0.744	.1202278	4.545383
OtherMale	688461.8	6.05e+08	0.02	0.988	0	.
3						
tenure	1.01786	.0118107	1.53	0.127	.9949731	1.041274
exper	1.022633	.0178585	1.28	0.200	.9882228	1.05824
pyperstudlep	1.050985	.0201998	2.59	0.010	1.012131	1.091331
pyperstude~s	1.009046	.0058035	1.57	0.117	.9977353	1.020485
pyperstudw~e	1.000147	.0004932	0.30	0.765	.999181	1.001114
pyperstudb~k	.9995646	.0007536	-0.58	0.563	.9980886	1.001043
pyperstudl~o	.9958157	.0012919	-3.23	0.001	.9932869	.998351
pyperstudm~e	.9949811	.0071583	-0.70	0.484	.9810496	1.009111
male	1.560896	.295673	2.35	0.019	1.0768	2.262627
age	1.000154	.015921	0.01	0.992	.9694314	1.031851
Asian	(omitted)					
Black	1.189544	.4685356	0.44	0.659	.5496806	2.574248
Latino	.8563709	.7539856	-0.18	0.860	.1524852	4.809459
Other	1.75e-06	.0013147	-0.02	0.986	0	.
AsianMale	(omitted)					
BlackMale	1.285588	.8521978	0.38	0.705	.3506329	4.713582
LatinoMale	1.225637	1.540003	0.16	0.871	.1044321	14.38434
OtherMale	1.190796	1181.092	0.00	1.000	0	.

```

-----> region = 13

```

```

Iteration 0: log likelihood = -1342.9199
Iteration 1: log likelihood = -1262.1826
Iteration 2: log likelihood = -1259.541
Iteration 3: log likelihood = -1259.3708
Iteration 4: log likelihood = -1259.3307
Iteration 5: log likelihood = -1259.3231
Iteration 6: log likelihood = -1259.3218
Iteration 7: log likelihood = -1259.3215

```

Iteration 8: log likelihood = -1259.3214  
 Iteration 9: log likelihood = -1259.3214

Multinomial logistic regression	Number of obs	=	1236
	LR chi2(36)	=	167.20
	Prob > chi2	=	0.0000
Log likelihood = -1259.3214	Pseudo R2	=	0.0623

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.028198	.0112792	2.53	0.011	1.006327	1.050544
	exper	.9742466	.0136058	-1.87	0.062	.9479414	1.001282
pyperstudlep		1.007283	.0114225	0.64	0.522	.9851419	1.029921
pyperstude~s		.9982655	.0059569	-0.29	0.771	.9866581	1.010009
pyperstudw~e		1.000028	.0002022	0.14	0.889	.999632	1.000425
pyperstudb~k		1.000844	.0006712	1.26	0.208	.9995296	1.002161
pyperstudl~o		1.00035	.0002971	1.18	0.239	.9997677	1.000932
pyperstudm~e		.9918378	.0070149	-1.16	0.247	.9781837	1.005682
	male	.88353	.1481777	-0.74	0.460	.6360123	1.227375
	age	.9817723	.0123194	-1.47	0.143	.9579213	1.006217
	Asian	1.627108	2.329707	0.34	0.734	.0983238	26.92613
	Black	1.005082	.3257386	0.02	0.988	.5325177	1.897006
	Latino	1.557984	.4019206	1.72	0.086	.9396681	2.583161
	Other	.4505796	.573588	-0.63	0.531	.0371699	5.461996
	AsianMale	1.92e+07	8.52e+10	0.00	0.997	0	.
	BlackMale	3.234263	1.763757	2.15	0.031	1.110671	9.418144
	LatinoMale	2.335135	1.019632	1.94	0.052	.9922856	5.495249
	OtherMale	1.152921	1.992789	0.08	0.934	.0389515	34.12523
3							
	tenure	1.008547	.0105825	0.81	0.417	.9880179	1.029504
	exper	1.014995	.0142262	1.06	0.288	.9874919	1.043265
pyperstudlep		1.010688	.0111026	0.97	0.333	.9891594	1.032684
pyperstude~s		1.004364	.0056827	0.77	0.442	.9932879	1.015564
pyperstudw~e		.9992603	.0002624	-2.82	0.005	.9987462	.9997746
pyperstudb~k		.9989306	.0008303	-1.29	0.198	.9973045	1.000559
pyperstudl~o		.9987316	.0004238	-2.99	0.003	.9979013	.9995626
pyperstudm~e		.9927845	.0054129	-1.33	0.184	.9822318	1.00345
	male	1.23183	.197392	1.30	0.193	.8998119	1.686358
	age	1.005845	.0129357	0.45	0.650	.9808088	1.031521
	Asian	1.797336	2.569694	0.41	0.682	.1090552	29.62186
	Black	.7564342	.2867873	-0.74	0.462	.3597946	1.590331
	Latino	1.02305	.2961469	0.08	0.937	.5800906	1.804255
	Other	3.88e-07	.0005923	-0.01	0.992	0	.
	AsianMale	.5645285	3875.686	-0.00	1.000	0	.
	BlackMale	2.201438	1.409068	1.23	0.218	.6278851	7.718496
	LatinoMale	2.058338	.9902233	1.50	0.133	.8017176	5.284597
	OtherMale	1.00555	2136.872	0.00	1.000	0	.

-----> region = 14

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -294.41039  
 Iteration 1: log likelihood = -262.39258  
 Iteration 2: log likelihood = -256.73447  
 Iteration 3: log likelihood = -254.63714  
 Iteration 4: log likelihood = -254.46036  
 Iteration 5: log likelihood = -254.43467  
 Iteration 6: log likelihood = -254.42865

```

Iteration 7:  log likelihood = -254.42716
Iteration 8:  log likelihood = -254.42687
Iteration 9:  log likelihood = -254.42683
Iteration 10: log likelihood = -254.42682

```

```

Multinomial logistic regression          Number of obs   =       311
                                         LR chi2(30)      =       79.97
                                         Prob > chi2      =      0.0000
Log likelihood = -254.42682             Pseudo R2       =      0.1358

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9960974	.0336139	-0.12	0.908	.9323468	1.064207
	exper	.973346	.0424778	-0.62	0.536	.8935524	1.060265
pyperstudlep		.8200588	.1254179	-1.30	0.195	.607665	1.106689
pyperstude~s		.965941	.0205964	-1.63	0.104	.9264046	1.007165
pyperstudw~e		.9975557	.0016836	-1.45	0.147	.9942614	1.000861
pyperstudb~k		1.011845	.0073074	1.63	0.103	.9976241	1.02627
pyperstudl~o		1.001738	.0016803	1.04	0.301	.99845	1.005036
pyperstudm~e		1.024645	.019394	1.29	0.198	.9873298	1.06337
	male	2.538642	1.164056	2.03	0.042	1.033465	6.236014
	age	.996235	.0383822	-0.10	0.922	.9237775	1.074376
	Asian	(omitted)					
	Black	2.22e-06	.0027456	-0.01	0.992	0	.
	Latino	1.833729	2.395651	0.46	0.643	.1416755	23.73426
	Other	3.65e-06	.0064548	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1.47e+12	2.33e+15	0.02	0.986	0	.
	LatinoMale	3.452185	5.503882	0.78	0.437	.1517078	78.55618
	OtherMale	(omitted)					
3							
	tenure	.9932372	.0192273	-0.35	0.726	.9562584	1.031646
	exper	1.033819	.0287351	1.20	0.231	.9790062	1.091702
pyperstudlep		1.045621	.043132	1.08	0.279	.9644107	1.133669
pyperstude~s		1.004838	.0128149	0.38	0.705	.9800328	1.030272
pyperstudw~e		.9994061	.0011345	-0.52	0.601	.997185	1.001632
pyperstudb~k		1.000266	.0052075	0.05	0.959	.9901113	1.010525
pyperstudl~o		.9970998	.0016388	-1.77	0.077	.9938929	1.000317
pyperstudm~e		.9953705	.0153751	-0.30	0.764	.9656874	1.025966
	male	2.661888	.7431033	3.51	0.000	1.540157	4.600602
	age	.9815822	.0250054	-0.73	0.466	.9337759	1.031836
	Asian	(omitted)					
	Black	6.378671	7.910188	1.49	0.135	.5612489	72.49448
	Latino	.4419111	.5233928	-0.69	0.491	.0433695	4.502826
	Other	.703632	1.037888	-0.24	0.812	.0390649	12.67374
	AsianMale	(omitted)					
	BlackMale	334115.5	3.30e+08	0.01	0.990	0	.
	LatinoMale	1.137026	1.979675	0.07	0.941	.0374752	34.49826
	OtherMale	(omitted)					

```

-----> region = 15

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: BlackMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0:  log likelihood = -311.33607
Iteration 1:  log likelihood = -280.38573
Iteration 2:  log likelihood = -275.11807
Iteration 3:  log likelihood = -274.96158

```

```

Iteration 4: log likelihood = -274.94033
Iteration 5: log likelihood = -274.93672
Iteration 6: log likelihood = -274.93589
Iteration 7: log likelihood = -274.93569
Iteration 8: log likelihood = -274.93565
Iteration 9: log likelihood = -274.93564

```

```

Multinomial logistic regression
Number of obs   =      313
LR chi2(28)     =      72.80
Prob > chi2     =      0.0000
Pseudo R2      =      0.1169

Log likelihood = -274.93564

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.029125	.0314811	0.94	0.348	.9692363	1.092714
	exper	1.010845	.0454056	0.24	0.810	.9256565	1.103873
pyperstudlep		.9987263	.024127	-0.05	0.958	.9525403	1.047152
pyperstude~s		.9827263	.0130559	-1.31	0.190	.9574675	1.008651
pyperstudw~e		1.000995	.0010137	0.98	0.326	.9990097	1.002983
pyperstudb~k		.9986891	.006008	-0.22	0.827	.9869828	1.010534
pyperstudl~o		1.000211	.0006305	0.34	0.738	.9989763	1.001448
pyperstudm~e		1.000742	.0210673	0.04	0.972	.9602909	1.042897
	male	1.44757	.6876963	0.78	0.436	.5705056	3.672988
	age	.9566197	.0384171	-1.10	0.269	.8842107	1.034958
	Asian	(omitted)					
	Black	6.29e-07	.0017337	-0.01	0.996	0	.
	Latino	10.05708	6.481349	3.58	0.000	2.84387	35.56593
	Other	8.610335	32910.38	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.4055327	.3226116	-1.13	0.257	.0852851	1.928317
	OtherMale	(omitted)					
3							
	tenure	.9972079	.0193774	-0.14	0.886	.9599431	1.035919
	exper	1.027777	.0275108	1.02	0.306	.975247	1.083137
pyperstudlep		1.010627	.0199533	0.54	0.592	.9722665	1.050502
pyperstude~s		1.007631	.0095326	0.80	0.422	.9891194	1.026489
pyperstudw~e		1.000054	.0010574	0.05	0.959	.997984	1.002129
pyperstudb~k		.9990122	.0065242	-0.15	0.880	.9863065	1.011882
pyperstudl~o		.997707	.0009274	-2.47	0.014	.9958909	.9995264
pyperstudm~e		.9845505	.0127763	-1.20	0.230	.9598253	1.009913
	male	1.377338	.3985301	1.11	0.269	.7811721	2.428477
	age	.9957547	.023803	-0.18	0.859	.9501778	1.043518
	Asian	(omitted)					
	Black	4.04e-07	.0007029	-0.01	0.993	0	.
	Latino	.9833915	.6388595	-0.03	0.979	.2752606	3.513249
	Other	3440362	6.24e+09	0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.388257	1.080733	0.42	0.673	.3018732	6.384325
	OtherMale	(omitted)					

```

-----> region = 16

```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -482.16848
Iteration 1: log likelihood = -424.70876
Iteration 2: log likelihood = -420.34052
Iteration 3: log likelihood = -419.85638
Iteration 4: log likelihood = -419.75741

```

```

Iteration 5: log likelihood = -419.73302
Iteration 6: log likelihood = -419.72821
Iteration 7: log likelihood = -419.72742
Iteration 8: log likelihood = -419.72733
Iteration 9: log likelihood = -419.72731

```

```

Multinomial logistic regression
Number of obs   =      454
LR chi2(34)     =     124.88
Prob > chi2     =     0.0000
Pseudo R2      =     0.1295

Log likelihood = -419.72731

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.069556	.0273716	2.63	0.009	1.017232	1.124572
exper	.9018406	.026927	-3.46	0.001	.8505793	.9561913
pyperstudlep	1.006216	.0163525	0.38	0.703	.9746708	1.038782
pyperstude~s	.9918204	.0083708	-0.97	0.330	.9755488	1.008363
pyperstudw~e	1.000623	.0004508	1.38	0.167	.9997395	1.001506
pyperstudb~k	1.001762	.0020644	0.85	0.393	.9977236	1.005816
pyperstudl~o	1.000568	.00064	0.89	0.375	.9993141	1.001823
pyperstudm~e	1.020268	.0113607	1.80	0.072	.9982427	1.042779
male	.8014609	.2370119	-0.75	0.454	.4489116	1.430882
age	1.023	.0237372	0.98	0.327	.9775179	1.070598
Asian	3.98e+07	1.99e+11	0.00	0.997	0	.
Black	.9274011	.8634384	-0.08	0.935	.1495455	5.751245
Latino	3.263981	2.04276	1.89	0.059	.9572465	11.12939
Other	.4116435	.5313848	-0.69	0.492	.0327882	5.168035
AsianMale	(omitted)					
BlackMale	3.28e-07	.0003871	-0.01	0.990	0	.
LatinoMale	.4197044	.4794863	-0.76	0.447	.0447197	3.939024
OtherMale	1.75e+07	2.40e+10	0.01	0.990	0	.
3						
tenure	1.054702	.0181432	3.10	0.002	1.019735	1.090868
exper	1.002118	.0239346	0.09	0.929	.9562882	1.050144
pyperstudlep	1.030711	.0152993	2.04	0.042	1.001157	1.061137
pyperstude~s	.9960228	.007284	-0.54	0.586	.9818482	1.010402
pyperstudw~e	.9996269	.0005214	-0.72	0.474	.9986054	1.000649
pyperstudb~k	.9967746	.0024726	-1.30	0.193	.9919402	1.001633
pyperstudl~o	.9983136	.0008463	-1.99	0.046	.9966563	.9999737
pyperstudm~e	1.004995	.0101495	0.49	0.622	.985298	1.025086
male	1.748977	.4368696	2.24	0.025	1.071929	2.853659
age	1.020761	.0231334	0.91	0.365	.976413	1.067124
Asian	2.015692	15323.86	0.00	1.000	0	.
Black	1.10e-06	.000918	-0.02	0.987	0	.
Latino	.2894611	.3437231	-1.04	0.296	.0282373	2.967269
Other	9.13e-07	.0008419	-0.02	0.988	0	.
AsianMale	(omitted)					
BlackMale	877951.6	7.32e+08	0.02	0.987	0	.
LatinoMale	4.13152	6.321736	0.93	0.354	.2059047	82.8998
OtherMale	8.41e+11	1.39e+15	0.02	0.987	0	.

```

-----> region = 17

```

```

Iteration 0: log likelihood = -470.14834
Iteration 1: log likelihood = -417.69016
Iteration 2: log likelihood = -413.82625
Iteration 3: log likelihood = -413.71099
Iteration 4: log likelihood = -413.68533
Iteration 5: log likelihood = -413.67956
Iteration 6: log likelihood = -413.67832

```

Iteration 7: log likelihood = -413.67812  
 Iteration 8: log likelihood = -413.67809

Multinomial logistic regression	Number of obs	=	469
	LR chi2(36)	=	112.94
	Prob > chi2	=	0.0000
Log likelihood = -413.67809	Pseudo R2	=	0.1201

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.997598	.0219126	-0.11	0.913	.9555614	1.041484
	exper	.9821437	.0323966	-0.55	0.585	.9206566	1.047737
pyperstudlep		1.029167	.0328029	0.90	0.367	.9668413	1.09551
pyperstude~s		.9961342	.0119923	-0.32	0.748	.9729048	1.019918
pyperstudw~e		1.001369	.0008087	1.69	0.090	.999785	1.002955
pyperstudb~k		1.004014	.0021407	1.88	0.060	.9998267	1.008218
pyperstudl~o		1.000555	.0007954	0.70	0.485	.9989975	1.002115
pyperstudm~e		.9999922	.0110932	-0.00	0.999	.9784846	1.021973
	male	3.7316	1.321619	3.72	0.000	1.863912	7.47076
	age	1.01958	.0297232	0.67	0.506	.9629571	1.079533
	Asian	6.27e-06	.0103432	-0.01	0.994	0	.
	Black	3.777958	3.784786	1.33	0.185	.5302944	26.91517
	Latino	1.01695	1.122948	0.02	0.988	.1167821	8.855705
	Other	1.950978	3993.773	0.00	1.000	0	.
	AsianMale	201109.1	5.29e+08	0.00	0.996	0	.
	BlackMale	.1777816	.283017	-1.08	0.278	.0078493	4.026664
	LatinoMale	2.075427	2.757502	0.55	0.583	.1535199	28.05759
	OtherMale	.9615172	1968.286	-0.00	1.000	0	.
3							
	tenure	1.008094	.0154995	0.52	0.600	.9781686	1.038935
	exper	1.024069	.0244234	1.00	0.319	.9773017	1.073075
pyperstudlep		1.031509	.0209022	1.53	0.126	.9913447	1.073301
pyperstude~s		1.00714	.0088127	0.81	0.416	.9900145	1.024561
pyperstudw~e		.9994141	.0008277	-0.71	0.479	.997793	1.001038
pyperstudb~k		.9974981	.0021854	-1.14	0.253	.9932241	1.001791
pyperstudl~o		.9990976	.0008955	-1.01	0.314	.9973439	1.000854
pyperstudm~e		.9837579	.0079909	-2.02	0.044	.9682201	.9995451
	male	2.998015	.7290746	4.51	0.000	1.861374	4.828741
	age	1.00187	.021726	0.09	0.931	.96018	1.04537
	Asian	1.68e-06	.0016522	-0.01	0.989	0	.
	Black	1.368783	1.301683	0.33	0.741	.2122562	8.826912
	Latino	1.393657	.7937354	0.58	0.560	.4564232	4.255439
	Other	1864366	1.86e+09	0.01	0.988	0	.
	AsianMale	1.73e+12	2.43e+15	0.02	0.984	0	.
	BlackMale	1.271815	1.712538	0.18	0.858	.0908354	17.80709
	LatinoMale	1.973462	1.64189	0.82	0.414	.3864048	10.07895
	OtherMale	1.01e-06	.0010097	-0.01	0.989	0	.

-----> region = 18

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -414.30151  
 Iteration 1: log likelihood = -366.18828  
 Iteration 2: log likelihood = -363.21752  
 Iteration 3: log likelihood = -363.15022  
 Iteration 4: log likelihood = -363.13427  
 Iteration 5: log likelihood = -363.13107  
 Iteration 6: log likelihood = -363.13039

```

Iteration 7:  log likelihood = -363.13023
Iteration 8:  log likelihood = -363.13019
Iteration 9:  log likelihood = -363.13019

```

```

Multinomial logistic regression      Number of obs   =      407
                                     LR chi2(30)        =     102.34
                                     Prob > chi2         =     0.0000
Log likelihood = -363.13019          Pseudo R2        =     0.1235

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9942873	.0234073	-0.24	0.808	.9494522	1.04124
	exper	.9831231	.0315913	-0.53	0.596	.9231149	1.047032
pyperstudlep		1.019621	.0184477	1.07	0.283	.9840973	1.056426
pyperstude~s		1.003975	.0145475	0.27	0.784	.9758636	1.032897
pyperstudw~e		1.000451	.0007241	0.62	0.533	.999033	1.001871
pyperstudb~k		.9998836	.002425	-0.05	0.962	.9951419	1.004648
pyperstudl~o		1.000947	.0007522	1.26	0.208	.999474	1.002422
pyperstudm~e		1.021021	.0128422	1.65	0.098	.9961589	1.046504
	male	1.866861	.6614984	1.76	0.078	.9321797	3.738732
	age	.9644714	.0269014	-1.30	0.195	.913161	1.018665
	Asian	(omitted)					
	Black	3.661946	5.382221	0.88	0.377	.2054191	65.28045
	Latino	1.419341	.8857496	0.56	0.575	.4177232	4.822641
	Other	4.01546	21970.87	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	1.55e-07	.0002717	-0.01	0.993	0	.
	LatinoMale	.8908447	.6816035	-0.15	0.880	.1988522	3.990926
	OtherMale	(omitted)					
3							
	tenure	.9973038	.0161495	-0.17	0.868	.9661485	1.029464
	exper	1.103505	.032522	3.34	0.001	1.041569	1.169124
pyperstudlep		.9894102	.0150728	-0.70	0.485	.9603047	1.019398
pyperstude~s		1.031166	.0109936	2.88	0.004	1.009843	1.05294
pyperstudw~e		.9998232	.0008374	-0.21	0.833	.9981833	1.001466
pyperstudb~k		.9986894	.0030556	-0.43	0.668	.9927185	1.004696
pyperstudl~o		.9981424	.0008359	-2.22	0.026	.9965053	.9997822
pyperstudm~e		.9882921	.0111326	-1.05	0.296	.9667116	1.010354
	male	1.663939	.4523413	1.87	0.061	.9766485	2.83489
	age	.9212222	.0252772	-2.99	0.003	.8729883	.972121
	Asian	(omitted)					
	Black	4.78073	5.779326	1.29	0.196	.4471922	51.10862
	Latino	1.714279	.8436173	1.10	0.273	.6534237	4.497467
	Other	2.10e+07	5.83e+10	0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	.3946226	.6860227	-0.53	0.593	.0130746	11.91061
	LatinoMale	.549913	.3516557	-0.94	0.350	.157026	1.925822
	OtherMale	(omitted)					

```

-----> region = 19

```

```

note: OtherMale omitted because of collinearity
Iteration 0:  log likelihood = -804.5742
Iteration 1:  log likelihood = -738.76377
Iteration 2:  log likelihood = -735.66455
Iteration 3:  log likelihood = -735.57798
Iteration 4:  log likelihood = -735.55773
Iteration 5:  log likelihood = -735.55377
Iteration 6:  log likelihood = -735.55312
Iteration 7:  log likelihood = -735.55296

```



Iteration 8: log likelihood = -735.55292

Multinomial logistic regression	Number of obs	=	750
	LR chi2(34)	=	138.04
	Prob > chi2	=	0.0000
Log likelihood = -735.55292	Pseudo R2	=	0.0858

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.001875	.013291	0.14	0.888	.9761612	1.028267
exper	1.056046	.0189696	3.04	0.002	1.019513	1.093888
pyperstudlep	1.014864	.0060426	2.48	0.013	1.003089	1.026776
pyperstude~s	.9966752	.0078584	-0.42	0.673	.9813915	1.012197
pyperstudw~e	1.000108	.0007695	0.14	0.888	.9986011	1.001617
pyperstudb~k	1.003347	.0020972	1.60	0.110	.9992451	1.007466
pyperstudl~o	.9999422	.0001885	-0.31	0.759	.9995729	1.000312
pyperstudm~e	.9909506	.0083553	-1.08	0.281	.9747091	1.007463
male	1.291283	.3492885	0.95	0.345	.7599307	2.194164
age	.9811853	.0149683	-1.25	0.213	.9522822	1.010966
Asian	908730.2	7.71e+08	0.02	0.987	0	.
Black	1.070164	.727154	0.10	0.921	.282538	4.053441
Latino	1.14438	.2878127	0.54	0.592	.6990247	1.873477
Other	2.14e-06	.001653	-0.02	0.987	0	.
AsianMale	6.52e-12	6.46e-09	-0.03	0.979	0	.
BlackMale	.6739221	.6856128	-0.39	0.698	.0917576	4.949681
LatinoMale	.9842516	.3518744	-0.04	0.965	.4884248	1.983419
OtherMale	(omitted)					
3						
tenure	1.033776	.0154234	2.23	0.026	1.003984	1.064451
exper	1.108792	.0256077	4.47	0.000	1.05972	1.160135
pyperstudlep	1.016875	.0066751	2.55	0.011	1.003875	1.030042
pyperstude~s	1.002763	.00898	0.31	0.758	.9853159	1.020519
pyperstudw~e	.999365	.0011202	-0.57	0.571	.9971717	1.001563
pyperstudb~k	1.002512	.0028553	0.88	0.378	.9969315	1.008124
pyperstudl~o	.9991872	.0002549	-3.19	0.001	.9986878	.999687
pyperstudm~e	.9980049	.0074693	-0.27	0.790	.9834722	1.012752
male	1.008875	.3365758	0.03	0.979	.5246431	1.940041
age	.9781964	.019415	-1.11	0.267	.9408745	1.016999
Asian	.9585	1352.215	-0.00	1.000	0	.
Black	.9687135	.7404068	-0.04	0.967	.2165734	4.332968
Latino	1.060659	.3070389	0.20	0.839	.6014095	1.8706
Other	2.85e-06	.0026052	-0.01	0.989	0	.
AsianMale	2.005456	2829.223	0.00	1.000	0	.
BlackMale	2.550443	2.771008	0.86	0.389	.3032501	21.45014
LatinoMale	1.268365	.5471226	0.55	0.582	.54459	2.954058
OtherMale	(omitted)					

-----> region = 20

Iteration 0: log likelihood = -2101.9227  
Iteration 1: log likelihood = -1951.7321  
Iteration 2: log likelihood = -1938.4313  
Iteration 3: log likelihood = -1937.9159  
Iteration 4: log likelihood = -1937.8163  
Iteration 5: log likelihood = -1937.7927  
Iteration 6: log likelihood = -1937.7875  
Iteration 7: log likelihood = -1937.7863  
Iteration 8: log likelihood = -1937.7861  
Iteration 9: log likelihood = -1937.7861

Multinomial logistic regression

Number of obs = 2110

LR chi2(36) = 328.27

Prob > chi2 = 0.0000

Pseudo R2 = 0.0781

Log likelihood = -1937.7861

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	.9943409	.0083219	-0.68	0.498	.9781632	1.010786
	exper	1.007961	.0111668	0.72	0.474	.9863108	1.030087
	pyperstudlep	.9949174	.0072908	-0.70	0.487	.9807297	1.00931
	pyperstude~s	.9981724	.0039291	-0.46	0.642	.9905012	1.005903
	pyperstudw~e	1.000018	.0002017	0.09	0.928	.9996229	1.000414
	pyperstudb~k	1.000339	.0004802	0.71	0.480	.999398	1.00128
	pyperstudl~o	1.000075	.0001236	0.61	0.541	.9998333	1.000318
	pyperstudm~e	.9982987	.0043142	-0.39	0.694	.9898788	1.00679
	male	1.488839	.2159342	2.74	0.006	1.120453	1.978345
	age	.9786924	.0093805	-2.25	0.025	.9604784	.9972517
	Asian	3.59e-07	.0003639	-0.01	0.988	0	.
	Black	1.306178	.4058354	0.86	0.390	.7104447	2.401456
	Latino	1.279398	.2118973	1.49	0.137	.9247565	1.770043
	Other	.5723525	.6315835	-0.51	0.613	.0658224	4.976837
	AsianMale	.5088989	5.16e+09	0.02	0.988	0	.
	BlackMale	.8610292	.3994963	-0.32	0.747	.3468019	2.137738
	LatinoMale	1.489433	.3606044	1.65	0.100	.9266968	2.393889
	OtherMale	1.929785	3.196087	0.40	0.691	.0751206	49.57456
3							
	tenure	1.024528	.0088175	2.82	0.005	1.007391	1.041956
	exper	1.083939	.0150893	5.79	0.000	1.054764	1.113921
	pyperstudlep	1.003706	.0067882	0.55	0.584	.9904893	1.017099
	pyperstude~s	.9992362	.00434	-0.18	0.860	.9907661	1.007779
	pyperstudw~e	.9987156	.0003544	-3.62	0.000	.9980212	.9994104
	pyperstudb~k	.9987413	.0006883	-1.83	0.068	.9973932	1.000091
	pyperstudl~o	.9986453	.000189	-7.16	0.000	.9982748	.9990159
	pyperstudm~e	.9911964	.0039752	-2.20	0.027	.9834357	.9990184
	male	1.729926	.2781366	3.41	0.001	1.262327	2.370736
	age	.9618645	.0119472	-3.13	0.002	.9387312	.9855678
	Asian	1.07e-06	.0010127	-0.01	0.988	0	.
	Black	1.473094	.4649813	1.23	0.220	.7935041	2.734713
	Latino	1.141777	.2032342	0.74	0.456	.8055069	1.618429
	Other	1.11e-06	.0009533	-0.02	0.987	0	.
	AsianMale	1.580973	3310.187	0.00	1.000	0	.
	BlackMale	.8710177	.4253689	-0.28	0.777	.3344533	2.268394
	LatinoMale	1.07634	.2888166	0.27	0.784	.6361279	1.821185
	OtherMale	2058041	1.76e+09	0.02	0.986	0	.

by psychooltype:

```
. bysort psychooltype: mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age A
> sian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert2000a==1,
rrr
```

```
-----> psychooltype =
```

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -390.91498

Iteration 1: log likelihood = -360.32551

Iteration 2: log likelihood = -359.15378

Iteration 3: log likelihood = -359.07322

Iteration 4: log likelihood = -359.05542

Iteration 5: log likelihood = -359.05172

```

Iteration 6: log likelihood = -359.05111
Iteration 7: log likelihood = -359.05097
Iteration 8: log likelihood = -359.05094
Iteration 9: log likelihood = -359.05093

```

```

Multinomial logistic regression
Log likelihood = -359.05093
Number of obs   =      366
LR chi2(34)     =      63.73
Prob > chi2     =      0.0015
Pseudo R2      =      0.0815

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.00606	.0224716	0.27	0.787	.9629671	1.051082
	exper	.987492	.0287193	-0.43	0.665	.9327774	1.045416
pypersistudlep		1.02699	.0121827	2.25	0.025	1.003388	1.051148
pypersistude~s		.995618	.0088306	-0.50	0.621	.97846	1.013077
pypersistudw~e		1.000113	.0003469	0.33	0.744	.9994335	1.000793
pypersistudb~k		1.000127	.0007597	0.17	0.867	.9986392	1.001617
pypersistudl~o		1.000453	.00036	1.26	0.208	.999748	1.001159
pypersistudm~e		1.012601	.0120087	1.06	0.291	.9893357	1.036413
	male	2.294448	.8608935	2.21	0.027	1.099768	4.786909
	age	1.006159	.0250946	0.25	0.806	.9581577	1.056566
	Asian	4.74e+07	1.73e+11	0.00	0.996	0	.
	Black	3.315975	2.017206	1.97	0.049	1.006468	10.92503
	Latino	1.111331	.5550974	0.21	0.833	.4175208	2.958074
	Other	1.454146	4257.713	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	2.184661	2.807836	0.61	0.543	.1759452	27.12631
	LatinoMale	.5333298	.4145542	-0.81	0.419	.1162416	2.446978
	OtherMale	1167176	3.61e+09	0.00	0.996	0	.
3							
	tenure	.9848259	.017727	-0.85	0.396	.9506875	1.02019
	exper	1.040653	.025424	1.63	0.103	.991997	1.091695
pypersistudlep		1.020034	.0109389	1.85	0.064	.9988177	1.0417
pypersistude~s		.9968391	.007447	-0.42	0.672	.9823495	1.011542
pypersistudw~e		.9990106	.0003828	-2.58	0.010	.9982606	.9997611
pypersistudb~k		.9998184	.0007247	-0.25	0.802	.9983989	1.00124
pypersistudl~o		.999798	.0003616	-0.56	0.576	.9990896	1.000507
pypersistudm~e		.9951692	.0104665	-0.46	0.645	.9748652	1.015896
	male	1.10126	.3404768	0.31	0.755	.6007913	2.018627
	age	1.005635	.021817	0.26	0.796	.9637708	1.049318
	Asian	1.634297	8554.04	0.00	1.000	0	.
	Black	.7433554	.4582643	-0.48	0.630	.2220481	2.488547
	Latino	.433066	.1896556	-1.91	0.056	.1835614	1.021708
	Other	1874750	3.28e+09	0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	6.663915	8.820391	1.43	0.152	.4978229	89.20394
	LatinoMale	1.663944	1.162695	0.73	0.466	.4230136	6.545205
	OtherMale	1.110442	2241.984	0.00	1.000	0	.

```
-----> pyschooltype = B
```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -735.10421
Iteration 1: log likelihood = -675.63087
Iteration 2: log likelihood = -669.21206
Iteration 3: log likelihood = -669.05932
Iteration 4: log likelihood = -669.02885
Iteration 5: log likelihood = -669.02216
Iteration 6: log likelihood = -669.02053

```

```
Iteration 7: log likelihood = -669.0202
Iteration 8: log likelihood = -669.02012
Iteration 9: log likelihood = -669.02011
```

```
Multinomial logistic regression      Number of obs   =      764
                                     LR chi2(34)        =     132.17
                                     Prob > chi2         =     0.0000
Log likelihood = -669.02011          Pseudo R2        =     0.0899
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.04447	.0183229	2.48	0.013	1.009168	1.081007
	exper	.9960698	.0222953	-0.18	0.860	.9533166	1.04074
	pyperstudlep	1.013275	.0093212	1.43	0.152	.9951696	1.03171
	pyperstude~s	1.001231	.0060493	0.20	0.839	.9894444	1.013158
	pyperstudw~e	1.001307	.0005019	2.61	0.009	1.000324	1.002291
	pyperstudb~k	1.00059	.0004228	1.39	0.163	.9997613	1.001419
	pyperstudl~o	1.000862	.00044	1.96	0.050	1	1.001725
	pyperstudm~e	1.013477	.0044892	3.02	0.003	1.004716	1.022314
	male	1.150037	.3641508	0.44	0.659	.6182796	2.139137
	age	.9445129	.0189071	-2.85	0.004	.9081732	.9823067
	Asian	1.155219	3110.897	0.00	1.000	0	.
	Black	2.090773	.8476418	1.82	0.069	.9445226	4.628084
	Latino	3.35402	1.822912	2.23	0.026	1.155945	9.731822
	Other	7.74e-06	.0039867	-0.02	0.982	0	.
	AsianMale	(omitted)					
	BlackMale	1.926742	1.010106	1.25	0.211	.6895667	5.383576
	LatinoMale	1.164813	.815365	0.22	0.827	.2954052	4.592977
	OtherMale	597782.6	3.08e+08	0.03	0.979	0	.
3							
	tenure	1.019512	.0119664	1.65	0.100	.9963264	1.043238
	exper	1.016422	.0156984	1.05	0.292	.9861147	1.047661
	pyperstudlep	1.000314	.0083853	0.04	0.970	.9840129	1.016884
	pyperstude~s	1.006431	.0046278	1.39	0.163	.9974009	1.015542
	pyperstudw~e	1.000441	.0005066	0.87	0.384	.9994489	1.001435
	pyperstudb~k	.998822	.0005239	-2.25	0.025	.9977958	.9998493
	pyperstudl~o	.9989705	.0005953	-1.73	0.084	.9978044	1.000138
	pyperstudm~e	1.001968	.0032619	0.60	0.546	.9955951	1.008382
	male	1.374329	.2747425	1.59	0.112	.9288095	2.033549
	age	.9815083	.0144053	-1.27	0.203	.9536766	1.010152
	Asian	2097276	2.58e+09	0.01	0.991	0	.
	Black	.6357858	.2656776	-1.08	0.278	.2802959	1.442131
	Latino	.9741818	.5638155	-0.05	0.964	.3133288	3.028863
	Other	1.327684	1.573527	0.24	0.811	.1301004	13.5491
	AsianMale	(omitted)					
	BlackMale	2.533404	1.309382	1.80	0.072	.9199483	6.976625
	LatinoMale	1.949038	1.363356	0.95	0.340	.4947698	7.677812
	OtherMale	2.192797	3.305204	0.52	0.602	.1142831	42.0741

```
-----> pyschooltype = E
```

```
Iteration 0: log likelihood = -9523.5672
Iteration 1: log likelihood = -8900.4786
Iteration 2: log likelihood = -8890.877
Iteration 3: log likelihood = -8890.8617
Iteration 4: log likelihood = -8890.8617
```

```
Multinomial logistic regression      Number of obs   =     8912
                                     LR chi2(36)        =    1265.41
                                     Prob > chi2         =     0.0000
```

Log likelihood = -8890.8617                      Pseudo R2                      =                      0.0664

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.002648	.0046812	0.57	0.571	.9935149	1.011865
	exper	.9864867	.0056769	-2.36	0.018	.9754228	.9976762
	pyperstudlep	1.01188	.0022172	5.39	0.000	1.007543	1.016235
	pyperstude~s	.9902936	.0020299	-4.76	0.000	.9863231	.9942802
	pyperstudw~e	1.001465	.0002207	6.64	0.000	1.001033	1.001898
	pyperstudb~k	1.00095	.0002311	4.11	0.000	1.000497	1.001403
	pyperstudl~o	1.000975	.0001775	5.50	0.000	1.000628	1.001323
	pyperstudm~e	1.016827	.0042237	4.02	0.000	1.008582	1.025139
	male	1.815686	.16612	6.52	0.000	1.51762	2.172294
	age	.9926265	.0046565	-1.58	0.115	.9835418	1.001795
	Asian	1.114436	.5118884	0.24	0.814	.4529775	2.741785
	Black	1.424104	.1544223	3.26	0.001	1.151442	1.761333
	Latino	1.14996	.1076634	1.49	0.136	.9571724	1.381578
	Other	.8468965	.3616541	-0.39	0.697	.3667227	1.955793
	AsianMale	.6786261	.5451373	-0.48	0.629	.1405608	3.2764
	BlackMale	.7272151	.1623206	-1.43	0.154	.4695345	1.126311
	LatinoMale	.7169698	.1219172	-1.96	0.050	.5137576	1.000561
	OtherMale	.1438303	.1637916	-1.70	0.089	.0154353	1.340247
3							
	tenure	1.010576	.00375	2.84	0.005	1.003252	1.017952
	exper	1.040813	.0053825	7.74	0.000	1.030317	1.051416
	pyperstudlep	1.005776	.0020008	2.90	0.004	1.001863	1.009705
	pyperstude~s	.9978531	.0018031	-1.19	0.234	.9943253	1.001394
	pyperstudw~e	.999681	.0002134	-1.49	0.135	.9992629	1.000099
	pyperstudb~k	.9987572	.0002291	-5.42	0.000	.9983082	.9992063
	pyperstudl~o	.9988958	.0001763	-6.26	0.000	.9985504	.9992413
	pyperstudm~e	1.002935	.0036499	0.81	0.421	.9958069	1.010115
	male	3.00092	.2274528	14.50	0.000	2.586653	3.481535
	age	1.00823	.0045282	1.83	0.068	.9993944	1.017145
	Asian	.5050656	.2966494	-1.16	0.245	.1597335	1.59698
	Black	1.065695	.11079	0.61	0.541	.8692438	1.306544
	Latino	1.197855	.1008976	2.14	0.032	1.015561	1.412871
	Other	.4803032	.2170843	-1.62	0.105	.1980591	1.164759
	AsianMale	.7188723	.6285652	-0.38	0.706	.1295332	3.989538
	BlackMale	.561942	.1148444	-2.82	0.005	.3764705	.8387878
	LatinoMale	.5284179	.0786737	-4.28	0.000	.3946808	.7074717
	OtherMale	.418951	.308321	-1.18	0.237	.0990225	1.772527

-----> pyschooltype = M

Iteration 0: log likelihood = -5411.0182  
Iteration 1: log likelihood = -5083.7705  
Iteration 2: log likelihood = -5073.2216  
Iteration 3: log likelihood = -5072.6541  
Iteration 4: log likelihood = -5072.5611  
Iteration 5: log likelihood = -5072.5399  
Iteration 6: log likelihood = -5072.5348  
Iteration 7: log likelihood = -5072.5338  
Iteration 8: log likelihood = -5072.5336  
Iteration 9: log likelihood = -5072.5335  
Iteration 10: log likelihood = -5072.5335

Multinomial logistic regression

Number of obs                      =                      5151  
LR chi2(36)                      =                      676.97  
Prob > chi2                      =                      0.0000  
Pseudo R2                      =                      0.0626

Log likelihood = -5072.5335

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005518	.0052337	1.06	0.290	.9953123	1.015828
	exper	.9945941	.0063372	-0.85	0.395	.9822507	1.007093
pyperstudlep		1.007374	.0036751	2.01	0.044	1.000197	1.014603
pyperstude~s		1.000005	.0025213	0.00	0.998	.9950755	1.004959
pyperstudw~e		1.000626	.0001607	3.90	0.000	1.000311	1.000941
pyperstudb~k		1.000825	.0001942	4.25	0.000	1.000445	1.001206
pyperstudl~o		1.000164	.0001362	1.20	0.229	.9998969	1.000431
pyperstudm~e		1.005122	.0036888	1.39	0.164	.9979176	1.012377
	male	1.727144	.14351	6.58	0.000	1.467579	2.032618
	age	.9731939	.0052394	-5.05	0.000	.9629788	.9835173
	Asian	.7671465	.5340871	-0.38	0.703	.1960072	3.00251
	Black	1.109064	.1480532	0.78	0.438	.8537415	1.440744
	Latino	1.880089	.2512218	4.72	0.000	1.446901	2.442969
	Other	.4539517	.2916644	-1.23	0.219	.1288581	1.599217
	AsianMale	.4123797	.54634	-0.67	0.504	.0307316	5.533615
	BlackMale	1.183828	.2300434	0.87	0.385	.8088752	1.732588
	LatinoMale	1.02766	.1798194	0.16	0.876	.7293005	1.448079
	OtherMale	2.647535	2.102873	1.23	0.220	.5581525	12.55829
3							
	tenure	1.004039	.0052064	0.78	0.437	.9938861	1.014295
	exper	1.048788	.0077384	6.46	0.000	1.03373	1.064065
pyperstudlep		1.003634	.0045528	0.80	0.424	.9947503	1.012597
pyperstude~s		.9980009	.0029409	-0.68	0.497	.9922535	1.003782
pyperstudw~e		.9992376	.0002089	-3.65	0.000	.9988282	.9996471
pyperstudb~k		.9988382	.0002789	-4.16	0.000	.9982918	.9993849
pyperstudl~o		.9986623	.0001892	-7.06	0.000	.9982915	.9990333
pyperstudm~e		.9878494	.0043188	-2.80	0.005	.9794209	.9963504
	male	2.249856	.202144	9.02	0.000	1.886585	2.683076
	age	.9792304	.0066358	-3.10	0.002	.9663106	.992323
	Asian	3.83e-07	.0004419	-0.01	0.990	0	.
	Black	.8974485	.1648802	-0.59	0.556	.6260743	1.286451
	Latino	1.98938	.3220813	4.25	0.000	1.448464	2.732295
	Other	2.96e-07	.0002897	-0.02	0.988	0	.
	AsianMale	5500742	6.35e+09	0.01	0.989	0	.
	BlackMale	1.309504	.3206035	1.10	0.271	.8104176	2.115948
	LatinoMale	.830908	.1685428	-0.91	0.361	.5583335	1.236551
	OtherMale	3209440	3.14e+09	0.02	0.988	0	.

-----> pyschooltype = S

Iteration 0: log likelihood = -6411.2402  
Iteration 1: log likelihood = -6038.2462  
Iteration 2: log likelihood = -6017.5025  
Iteration 3: log likelihood = -6017.3854  
Iteration 4: log likelihood = -6017.3854

Multinomial logistic regression	Number of obs	=	6683
	LR chi2(36)	=	787.71
	Prob > chi2	=	0.0000
Log likelihood = -6017.3854	Pseudo R2	=	0.0614

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						

tenure		.998013	.0040935	-0.48	0.628	.9900221	1.006068
exper		.9984896	.005297	-0.28	0.776	.9881614	1.008926
pyperstudlep		1.018113	.0038675	4.73	0.000	1.010561	1.025721
pyperstude~s		.995832	.0021726	-1.91	0.056	.9915829	1.000099
pyperstudw~e		1.000198	.0000615	3.21	0.001	1.000077	1.000318
pyperstudb~k		1.000311	.0000951	3.27	0.001	1.000124	1.000497
pyperstudl~o		1.000083	.0000605	1.37	0.169	.9999646	1.000202
pyperstudm~e		1.000015	.0024388	0.01	0.995	.9952461	1.004806
male		1.238845	.0887255	2.99	0.003	1.0766	1.425541
age		.9764511	.004805	-4.84	0.000	.9670787	.9859144
Asian		.73212	.4886041	-0.47	0.640	.1979271	2.708066
Black		1.589258	.2146953	3.43	0.001	1.219563	2.071021
Latino		1.511835	.1946314	3.21	0.001	1.174687	1.945749
Other		.7210134	.3696938	-0.64	0.524	.2639348	1.969654
AsianMale		2.125699	1.778579	0.90	0.367	.4123808	10.95734
BlackMale		1.530316	.2727212	2.39	0.017	1.079161	2.17008
LatinoMale		1.439534	.2261666	2.32	0.020	1.058008	1.958642
OtherMale		2.79651	1.818786	1.58	0.114	.7816532	10.00503
-----							
3							
tenure		.9949092	.004729	-1.07	0.283	.9856835	1.004221
exper		1.037573	.0072189	5.30	0.000	1.02352	1.051819
pyperstudlep		.9998563	.0053602	-0.03	0.979	.9894055	1.010418
pyperstude~s		.9951799	.0025502	-1.89	0.059	.990194	1.000191
pyperstudw~e		.9992241	.0000949	-8.18	0.000	.9990383	.9994101
pyperstudb~k		.9992785	.0001638	-4.40	0.000	.9989575	.9995997
pyperstudl~o		.9993807	.0000918	-6.75	0.000	.9992008	.9995606
pyperstudm~e		1.006763	.0021219	3.20	0.001	1.002613	1.010931
male		2.181674	.2100151	8.10	0.000	1.806552	2.634689
age		.9883409	.0067579	-1.72	0.086	.9751841	1.001675
Asian		1.114512	1.180523	0.10	0.918	.1397875	8.885902
Black		1.124239	.243572	0.54	0.589	.735261	1.718999
Latino		1.425164	.2756529	1.83	0.067	.9754993	2.082107
Other		.3782695	.3951543	-0.93	0.352	.0488208	2.930879
AsianMale		.3337884	.5004743	-0.73	0.464	.0176685	6.30583
BlackMale		.8975854	.2493346	-0.39	0.697	.5207459	1.547126
LatinoMale		.9796203	.2212871	-0.09	0.927	.6291862	1.525233
OtherMale		3.614169	4.155642	1.12	0.264	.3795598	34.41411
-----							

### 2000-01

```
. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino Oth
> er AsianMale BlackMale LatinoMale OtherMale if validcert2001a=1, rrr
```

```
Iteration 0: log likelihood = -23369.36
Iteration 1: log likelihood = -22278.242
Iteration 2: log likelihood = -22261.688
Iteration 3: log likelihood = -22261.665
Iteration 4: log likelihood = -22261.665
```

```
Multinomial logistic regression      Number of obs   =      21988
                                     LR chi2(36)      =      2215.39
                                     Prob > chi2      =      0.0000
Log likelihood = -22261.665          Pseudo R2       =      0.0474
```

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.009688	.0025434	3.83	0.000	1.004716 1.014685
exper		.9952566	.0033398	-1.42	0.157	.9887323 1.001824
pyperstudlep		1.009308	.0014685	6.37	0.000	1.006434 1.01219
pyperstude~s		.9905205	.0012209	-7.73	0.000	.9881304 .9929163

pyperstudw~e		.9780649	.0040851	-5.31	0.000	.9700909	.9861044
pyperstudb~k		.9836431	.0044373	-3.66	0.000	.9749845	.9923787
pyperstudl~o		.9815405	.0042476	-4.31	0.000	.9732506	.989901
pyperstudm~e		.9995711	.0014267	-0.30	0.764	.9967788	1.002371
male		1.27822	.0538767	5.82	0.000	1.176868	1.388301
age		.9703606	.0028043	-10.41	0.000	.9648797	.9758726
Asian		.6726266	.2176616	-1.23	0.220	.3567186	1.268301
Black		1.309104	.0894768	3.94	0.000	1.144972	1.496764
Latino		1.297438	.0820778	4.12	0.000	1.146142	1.468706
Other		.7654702	.1996542	-1.02	0.306	.4591062	1.276273
AsianMale		1.537636	.7376455	0.90	0.370	.6004951	3.937294
BlackMale		1.487578	.1543072	3.83	0.000	1.213903	1.822952
LatinoMale		1.385976	.1214885	3.72	0.000	1.167194	1.645766
OtherMale		1.523262	.5684939	1.13	0.259	.7329915	3.165556

3							
tenure		1.011401	.0023294	4.92	0.000	1.006846	1.015977
exper		1.055779	.0037544	15.26	0.000	1.048446	1.063163
pyperstudlep		1.011762	.001475	8.02	0.000	1.008875	1.014657
pyperstude~s		1.019707	.0012633	15.75	0.000	1.017234	1.022186
pyperstudw~e		1.00794	.0049073	1.62	0.104	.9983679	1.017604
pyperstudb~k		.9868559	.0051691	-2.53	0.012	.9767765	.9970392
pyperstudl~o		.9846089	.004955	-3.08	0.002	.9749451	.9943685
pyperstudm~e		1.006572	.0013943	4.73	0.000	1.003843	1.009309
male		1.454919	.0596426	9.15	0.000	1.342595	1.576641
age		.97295	.0031424	-8.49	0.000	.9668104	.9791285
Asian		.3771337	.1714368	-2.15	0.032	.154724	.9192486
Black		.9265339	.0710373	-1.00	0.320	.7972595	1.07677
Latino		1.180502	.077738	2.52	0.012	1.037561	1.343135
Other		.5168952	.1636722	-2.08	0.037	.2778906	.9614597
AsianMale		1.45366	.9287767	0.59	0.558	.4155377	5.085282
BlackMale		1.259218	.1484466	1.96	0.051	.9994354	1.586527
LatinoMale		1.057613	.0975545	0.61	0.544	.8826971	1.26719
OtherMale		1.875162	.7972935	1.48	0.139	.8149284	4.314775

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=21988)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		29.309	2	0.000
exper		289.194	2	0.000
pyperstudlep		77.827	2	0.000
pyperstude~s		452.631	2	0.000
pyperstudw~e		42.605	2	0.000
pyperstudb~k		15.109	2	0.001
pyperstudl~o		21.334	2	0.000
pyperstudm~e		25.664	2	0.000
male		92.257	2	0.000
age		142.332	2	0.000
Asian		6.042	2	0.049
Black		21.056	2	0.000
Latino		18.336	2	0.000
Other		5.083	2	0.079
AsianMale		0.954	2	0.621
BlackMale		15.012	2	0.001
LatinoMale		14.752	2	0.001
OtherMale		2.778	2	0.249

\*\*\*\* Wald tests for independent variables (N=21988)

Ho: All coefficients associated with given variable(s) are 0.



	chi2	df	P>chi2
tenure	29.283	2	0.000
exper	272.054	2	0.000
pyperstudlep	76.661	2	0.000
pyperstude~s	435.602	2	0.000
pyperstudw~e	43.162	2	0.000
pyperstudb~k	14.994	2	0.001
pyperstudl~o	21.125	2	0.000
pyperstudm~e	26.632	2	0.000
male	91.933	2	0.000
age	140.578	2	0.000
Asian	5.322	2	0.070
Black	21.303	2	0.000
Latino	18.354	2	0.000
Other	4.691	2	0.096
AsianMale	0.949	2	0.622
BlackMale	14.907	2	0.001
LatinoMale	14.744	2	0.001
OtherMale	2.736	2	0.255

\*\*\*\* Hausman tests of IIA assumption (N=21988)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-33.805	17	---	---
3	-20.676	18	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=21988)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-5045.344	-5033.460	23.767	19	0.205	for Ho
3	-5087.934	-5079.017	17.833	19	0.534	for Ho

\*\*\*\* Wald tests for combining alternatives (N=21988)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2-	3   1448.522	18	0.000
2-	1   642.742	18	0.000
3-	1   1065.756	18	0.000

\*\*\*\* LR tests for combining alternatives (N=21988)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2-	3   1625.068	18	0.000
2-	1   670.060	18	0.000
3-	1   1168.255	18	0.000

by region:

```
. bysort region: mlogit admin tenure exper pyperstudlep pyperstuddecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian B
> lack Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert2001a==1, rrr
```

-----> region = 1

```
Iteration 0: log likelihood = -1574.4108
Iteration 1: log likelihood = -1472.908
Iteration 2: log likelihood = -1469.1747
Iteration 3: log likelihood = -1468.8221
Iteration 4: log likelihood = -1468.7533
Iteration 5: log likelihood = -1468.7373
Iteration 6: log likelihood = -1468.7339
Iteration 7: log likelihood = -1468.7334
Iteration 8: log likelihood = -1468.7334
Iteration 9: log likelihood = -1468.7334
```

```
Multinomial logistic regression      Number of obs   =      1470
                                     LR chi2(36)        =      211.35
                                     Prob > chi2         =      0.0000
Log likelihood = -1468.7334          Pseudo R2        =      0.0671
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.023488	.0100825	2.36	0.018	1.003916	1.043441
	exper	1.030288	.0151947	2.02	0.043	1.000933	1.060504
pyperstudlep		.9990854	.0035941	-0.25	0.799	.9920658	1.006155
pyperstude~s		1.004474	.006533	0.69	0.493	.9917503	1.01736
pyperstudw~e		1.005844	.1035565	0.06	0.955	.8220446	1.230739
pyperstudb~k		.6133731	.1549219	-1.94	0.053	.3738812	1.006273
pyperstudl~o		.9859971	.0953814	-0.15	0.884	.8157063	1.191839
pyperstudm~e		.9922899	.0078142	-0.98	0.326	.977092	1.007724
	male	2.15929	.6093682	2.73	0.006	1.241925	3.754279
	age	.9482526	.0123659	-4.07	0.000	.924323	.9728017
	Asian	1.549351	2.225422	0.30	0.761	.0927941	25.86898
	Black	1.85e+07	5.05e+10	0.01	0.995	0	.
	Latino	1.119781	.248325	0.51	0.610	.7250511	1.729407
	Other	6.78e-07	.0007887	-0.01	0.990	0	.
	AsianMale	6.82e-08	.0001675	-0.01	0.995	0	.
	BlackMale	5.51e-08	.0001509	-0.01	0.995	0	.
	LatinoMale	.8840172	.2788466	-0.39	0.696	.4763928	1.640424
	OtherMale	1.56e+12	2.71e+15	0.02	0.987	0	.
3							
	tenure	1.027419	.0100858	2.76	0.006	1.00784	1.047378
	exper	1.079134	.0182163	4.51	0.000	1.044015	1.115435
pyperstudlep		1.014616	.0037192	3.96	0.000	1.007352	1.021931
pyperstude~s		1.016736	.0077729	2.17	0.030	1.001615	1.032085
pyperstudw~e		.9112722	.096719	-0.88	0.381	.7401244	1.121996
pyperstudb~k		.799723	.1320374	-1.35	0.176	.5786326	1.10529
pyperstudl~o		.8865613	.087828	-1.22	0.224	.7301023	1.076549
pyperstudm~e		1.013659	.0064501	2.13	0.033	1.001095	1.02638
	male	.6432478	.2374864	-1.20	0.232	.3119719	1.326298
	age	.9629175	.0149414	-2.44	0.015	.9340737	.992652
	Asian	1.01e-06	.0015304	-0.01	0.993	0	.
	Black	3.151721	14248.87	0.00	1.000	0	.
	Latino	1.094447	.2507079	0.39	0.694	.6985665	1.714675
	Other	.9688649	1.233592	-0.02	0.980	.0798885	11.75011
	AsianMale	1.046811	3204.489	0.00	1.000	0	.
	BlackMale	2.91e-07	.0013844	-0.00	0.997	0	.
	LatinoMale	2.301151	.9174432	2.09	0.037	1.053369	5.027009

```

OtherMale |      4651504      6.01e+09      0.01      0.991      0      .
-----

```

```

-----> region = 2

```

```

Iteration 0: log likelihood = -833.90795
Iteration 1: log likelihood = -786.48705
Iteration 2: log likelihood = -784.9632
Iteration 3: log likelihood = -784.76288
Iteration 4: log likelihood = -784.73044
Iteration 5: log likelihood = -784.72709
Iteration 6: log likelihood = -784.72641
Iteration 7: log likelihood = -784.72626
Iteration 8: log likelihood = -784.72623
Iteration 9: log likelihood = -784.72622

```

```

Multinomial logistic regression
Number of obs      =      848
LR chi2(36)        =      98.36
Prob > chi2        =      0.0000
Pseudo R2          =      0.0590
Log likelihood = -784.72622

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.011987	.0153686	0.78	0.433	.9823091	1.042562
exper	.9608382	.0183333	-2.09	0.036	.9255691	.9974512
pyperstudlep	1.005142	.0166479	0.31	0.757	.9730367	1.038307
pyperstude~s	1.004864	.0068178	0.72	0.474	.99159	1.018316
pyperstudw~e	.9380812	.066205	-0.91	0.365	.8168964	1.077243
pyperstudb~k	.9337054	.0686961	-0.93	0.351	.808321	1.078539
pyperstudl~o	.9332578	.062625	-1.03	0.303	.8182441	1.064438
pyperstudm~e	.9911148	.0077119	-1.15	0.251	.9761144	1.006346
male	1.309131	.3640675	0.97	0.333	.7590412	2.257881
age	.9996248	.0156955	-0.02	0.981	.9693307	1.030866
Asian	1.05e+08	5.37e+11	0.00	0.997	0	.
Black	4.983348	3.589808	2.23	0.026	1.214365	20.45
Latino	1.127212	.3036835	0.44	0.657	.6647864	1.9113
Other	4.40e-07	.0008838	-0.01	0.994	0	.
AsianMale	4.28e-15	2.50e-11	-0.01	0.995	0	.
BlackMale	.7712652	.7808674	-0.26	0.798	.1060241	5.610515
LatinoMale	1.514214	.5896989	1.07	0.287	.7058191	3.248487
OtherMale	2814563	5.65e+09	0.01	0.994	0	.
3						
tenure	.9926645	.0116829	-0.63	0.532	.9700285	1.015829
exper	1.045616	.0182616	2.55	0.011	1.01043	1.082028
pyperstudlep	.9977641	.0137862	-0.16	0.871	.9711062	1.025154
pyperstude~s	1.03522	.0068686	5.22	0.000	1.021845	1.04877
pyperstudw~e	1.152886	.0836474	1.96	0.050	1.000063	1.329061
pyperstudb~k	1.099511	.0833304	1.25	0.211	.9477381	1.27559
pyperstudl~o	1.121216	.0777891	1.65	0.099	.9786638	1.284532
pyperstudm~e	1.002787	.0071228	0.39	0.695	.9889234	1.016845
male	1.263405	.3180438	0.93	0.353	.7713745	2.069285
age	.9876146	.0159639	-0.77	0.441	.9568165	1.019404
Asian	1.809009	16637.42	0.00	1.000	0	.
Black	1.476321	1.089908	0.53	0.598	.3473546	6.274633
Latino	.8800311	.2094046	-0.54	0.591	.5520183	1.402951
Other	7.35e-07	.0012286	-0.01	0.993	0	.
AsianMale	4.02e-07	.0038229	-0.00	0.999	0	.
BlackMale	2.300937	2.336153	0.82	0.412	.3145399	16.83193
LatinoMale	1.507805	.5346462	1.16	0.247	.7525255	3.021128
OtherMale	1.47802	2911.259	0.00	1.000	0	.

-----> region = 3

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 note: OtherMale omitted because of collinearity  
 Iteration 0: log likelihood = -336.91318  
 Iteration 1: log likelihood = -304.93181  
 Iteration 2: log likelihood = -303.77546  
 Iteration 3: log likelihood = -303.72512  
 Iteration 4: log likelihood = -303.71365  
 Iteration 5: log likelihood = -303.71136  
 Iteration 6: log likelihood = -303.71098  
 Iteration 7: log likelihood = -303.71089  
 Iteration 8: log likelihood = -303.71087

Multinomial logistic regression	Number of obs	=	321
	LR chi2(30)	=	66.40
	Prob > chi2	=	0.0001
Log likelihood = -303.71087	Pseudo R2	=	0.0985

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.02127	.0225025	0.96	0.339	.9781043	1.06634
	exper	.9682882	.0304178	-1.03	0.305	.9104686	1.02978
pyperstudlep		.9070923	.0587772	-1.50	0.132	.7989063	1.029929
pyperstude~s		.9802994	.0139489	-1.40	0.162	.9533377	1.008023
pyperstudw~e		.988079	.0481242	-0.25	0.806	.8981194	1.087049
pyperstudb~k		1.003612	.0502333	0.07	0.943	.9098315	1.107058
pyperstudl~o		1.04673	.0532337	0.90	0.369	.9474256	1.156444
pyperstudm~e		.9813758	.017953	-1.03	0.304	.9468118	1.017201
	male	1.173793	.4343339	0.43	0.665	.5683618	2.424143
	age	.985005	.0280491	-0.53	0.596	.9315358	1.041543
	Asian	(omitted)					
	Black	1.656457	1.308576	0.64	0.523	.352163	7.791419
	Latino	1.869403	1.308833	0.89	0.372	.4739664	7.37324
	Other	384152.2	1.69e+08	0.03	0.977	0	.
	AsianMale	(omitted)					
	BlackMale	.4054857	.7115523	-0.51	0.607	.0130099	12.63795
	LatinoMale	.702661	.7401092	-0.34	0.738	.089162	5.537478
	OtherMale	(omitted)					
3							
	tenure	.9891993	.0173073	-0.62	0.535	.9558526	1.023709
	exper	1.007737	.0266575	0.29	0.771	.9568206	1.061363
pyperstudlep		1.050552	.0427655	1.21	0.226	.96999	1.137805
pyperstude~s		1.027296	.0120667	2.29	0.022	1.003916	1.051221
pyperstudw~e		1.102124	.0505853	2.12	0.034	1.007307	1.205866
pyperstudb~k		1.05523	.0496202	1.14	0.253	.9623228	1.157106
pyperstudl~o		1.090463	.0516347	1.83	0.067	.9938154	1.19651
pyperstudm~e		1.015792	.013062	1.22	0.223	.990511	1.041719
	male	2.829486	.908576	3.24	0.001	1.50792	5.309294
	age	.9995256	.0249156	-0.02	0.985	.9518656	1.049572
	Asian	(omitted)					
	Black	1.631649	1.216581	0.66	0.511	.3784065	7.0355
	Latino	1.928669	1.279051	0.99	0.322	.525726	7.075477
	Other	451672.8	1.99e+08	0.03	0.976	0	.
	AsianMale	(omitted)					
	BlackMale	.5548062	.840374	-0.39	0.697	.0284979	10.80115
	LatinoMale	.916829	.8580432	-0.09	0.926	.1464417	5.740002
	OtherMale	(omitted)					

-----> region = 4

Iteration 0: log likelihood = -4667.8416  
 Iteration 1: log likelihood = -4427.345  
 Iteration 2: log likelihood = -4419.8768  
 Iteration 3: log likelihood = -4419.8412  
 Iteration 4: log likelihood = -4419.8411

Multinomial logistic regression	Number of obs	=	4480
	LR chi2(36)	=	496.00
	Prob > chi2	=	0.0000
Log likelihood = -4419.8411	Pseudo R2	=	0.0531

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.015787	.0055254	2.88	0.004	1.005015	1.026674
	exper	1.020233	.0077407	2.64	0.008	1.005173	1.035517
pyperstudlep		1.001531	.0039139	0.39	0.695	.9938894	1.009232
pyperstude~s		.995937	.0033631	-1.21	0.228	.9893673	1.00255
pyperstudw~e		.9915925	.0066838	-1.25	0.210	.9785787	1.004779
pyperstudb~k		.9893541	.0078805	-1.34	0.179	.9740284	1.004921
pyperstudl~o		.9933673	.0076835	-0.86	0.390	.9784215	1.008541
pyperstudm~e		.9999435	.0034562	-0.02	0.987	.9931923	1.006741
	male	1.11593	.1032909	1.19	0.236	.9307852	1.337903
	age	.9535549	.0061036	-7.43	0.000	.9416667	.9655932
	Asian	.4608473	.2684883	-1.33	0.184	.1471124	1.443659
	Black	.9674584	.1103564	-0.29	0.772	.7736369	1.209839
	Latino	1.261843	.1937464	1.51	0.130	.9339199	1.704907
	Other	.2318582	.1774112	-1.91	0.056	.0517497	1.038813
	AsianMale	1.423558	1.286493	0.39	0.696	.2421757	8.36797
	BlackMale	1.56712	.2729373	2.58	0.010	1.113917	2.204711
	LatinoMale	1.679796	.4381572	1.99	0.047	1.007464	2.800809
	OtherMale	2.728872	3.116869	0.88	0.379	.2909083	25.59826
3							
	tenure	1.033208	.0063062	5.35	0.000	1.020921	1.045642
	exper	1.068791	.0102611	6.93	0.000	1.048868	1.089093
pyperstudlep		1.01856	.0046327	4.04	0.000	1.00952	1.02768
pyperstude~s		1.025025	.0041729	6.07	0.000	1.016879	1.033237
pyperstudw~e		.9987706	.0084109	-0.15	0.884	.9824207	1.015392
pyperstudb~k		.9783602	.0098212	-2.18	0.029	.9592992	.9978
pyperstudl~o		.9670824	.0092979	-3.48	0.000	.9490296	.9854787
pyperstudm~e		1.004453	.004016	1.11	0.266	.9966128	1.012355
	male	1.1527	.1247231	1.31	0.189	.9324282	1.425006
	age	.9642995	.0080418	-4.36	0.000	.9486659	.9801906
	Asian	.1575632	.169301	-1.72	0.085	.0191797	1.294397
	Black	.5346666	.077715	-4.31	0.000	.4021231	.7108977
	Latino	1.172552	.2068495	0.90	0.367	.8297971	1.656885
	Other	.5806171	.3853478	-0.82	0.413	.1581098	2.132166
	AsianMale	1.399553	2.133245	0.22	0.825	.0705601	27.75997
	BlackMale	1.380481	.3051479	1.46	0.145	.8951111	2.129041
	LatinoMale	1.402285	.4429409	1.07	0.284	.7550345	2.60439
	OtherMale	2.621197	2.830077	0.89	0.372	.3158399	21.75367

-----> region = 5

note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -570.93565  
 Iteration 1: log likelihood = -496.8993

```

Iteration 2: log likelihood = -493.47733
Iteration 3: log likelihood = -493.26982
Iteration 4: log likelihood = -493.23023
Iteration 5: log likelihood = -493.22178
Iteration 6: log likelihood = -493.21985
Iteration 7: log likelihood = -493.21938
Iteration 8: log likelihood = -493.21929
Iteration 9: log likelihood = -493.21927

```

```

Multinomial logistic regression
Number of obs   =      579
LR chi2(34)     =     155.43
Prob > chi2     =     0.0000
Pseudo R2      =     0.1361

Log likelihood = -493.21927

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.035747	.0190931	1.91	0.057	.9989933	1.073853
exper	.9849576	.0258898	-0.58	0.564	.9354995	1.037031
pyperstudlep	.9394501	.0336517	-1.74	0.081	.8757561	1.007777
pyperstude~s	1.01227	.0086617	1.43	0.154	.9954352	1.02939
pyperstudw~e	1.00491	.046098	0.11	0.915	.9185021	1.099446
pyperstudb~k	.9904236	.0463447	-0.21	0.837	.9036306	1.085553
pyperstudl~o	1.059231	.0659136	0.92	0.355	.9376099	1.196627
pyperstudm~e	1.017552	.0117884	1.50	0.133	.9947075	1.040921
male	3.563892	1.027999	4.41	0.000	2.024869	6.272667
age	.953944	.0201267	-2.23	0.025	.9153009	.9942187
Asian	1.69e-06	.0030397	-0.01	0.994	0	.
Black	2.564351	1.015634	2.38	0.017	1.179917	5.573185
Latino	3164606	2.88e+09	0.02	0.987	0	.
Other	1.04e-06	.0018717	-0.01	0.994	0	.
AsianMale	(omitted)					
BlackMale	.5984531	.3048636	-1.01	0.314	.2205015	1.624234
LatinoMale	7.56e-07	.0006884	-0.02	0.988	0	.
OtherMale	1.07e+13	3.52e+16	0.01	0.993	0	.
3						
tenure	1.002033	.0151009	0.13	0.893	.972869	1.032072
exper	1.095197	.0290156	3.43	0.001	1.039778	1.153568
pyperstudlep	.9930443	.0369656	-0.19	0.851	.9231728	1.068204
pyperstude~s	1.041077	.0090909	4.61	0.000	1.023411	1.059048
pyperstudw~e	.9722404	.0416402	-0.66	0.511	.8939586	1.057377
pyperstudb~k	.9389722	.0409702	-1.44	0.149	.8620098	1.022806
pyperstudl~o	.9439343	.0542673	-1.00	0.316	.8433458	1.05652
pyperstudm~e	1.033074	.011388	2.95	0.003	1.010993	1.055637
male	4.63409	1.276004	5.57	0.000	2.701374	7.949578
age	.9584645	.0230278	-1.77	0.077	.914377	1.004678
Asian	2.13e-06	.003693	-0.01	0.994	0	.
Black	2.065119	.9077581	1.65	0.099	.8725395	4.887707
Latino	8186705	7.45e+09	0.02	0.986	0	.
Other	6.77e-06	.0117123	-0.01	0.995	0	.
AsianMale	(omitted)					
BlackMale	.619684	.3471985	-0.85	0.393	.20666	1.858165
LatinoMale	6.96e-14	8.46e-11	-0.02	0.980	0	.
OtherMale	147449.5	8.03e+08	0.00	0.998	0	.

```
-----> region = 6
```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -769.3428
Iteration 1: log likelihood = -709.12091
Iteration 2: log likelihood = -708.11993

```

```

Iteration 3: log likelihood = -708.06557
Iteration 4: log likelihood = -708.0564
Iteration 5: log likelihood = -708.05432
Iteration 6: log likelihood = -708.05382
Iteration 7: log likelihood = -708.05372
Iteration 8: log likelihood = -708.05369

```

```

Multinomial logistic regression      Number of obs   =      711
                                     LR chi2(34)        =     122.58
                                     Prob > chi2         =      0.0000
Log likelihood = -708.05369          Pseudo R2        =      0.0797

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9592494	.013835	-2.88	0.004	.932513	.9867523
	exper	.9913832	.0170598	-0.50	0.615	.9585041	1.02539
pyperstudlep		1.055822	.0197575	2.90	0.004	1.0178	1.095265
pyperstude~s		.9989417	.0076854	-0.14	0.891	.9839916	1.014119
pyperstudw~e		.9572061	.0476469	-0.88	0.380	.8682308	1.055299
pyperstudb~k		.9678914	.0499999	-0.63	0.528	.8746912	1.071022
pyperstudl~o		.9409584	.0467373	-1.23	0.220	.8536725	1.037169
pyperstudm~e		1.002064	.0089496	0.23	0.817	.9846755	1.019759
	male	1.247554	.2560853	1.08	0.281	.8343211	1.865458
	age	.9792097	.0148446	-1.39	0.166	.9505429	1.008741
	Asian	5.40e-06	.0044918	-0.01	0.988	0	.
	Black	1.816101	1.014218	1.07	0.285	.6078248	5.42627
	Latino	.3372462	.2949049	-1.24	0.214	.0607595	1.871889
	Other	1.624485	2.502141	0.32	0.753	.0793666	33.25015
	AsianMale	(omitted)					
	BlackMale	1.360762	1.088275	0.39	0.700	.283812	6.524299
	LatinoMale	25.71782	36.04322	2.32	0.021	1.649246	401.0356
	OtherMale	1.48e-06	.0012289	-0.02	0.987	0	.
3							
	tenure	.9627982	.0127023	-2.87	0.004	.9382212	.9880189
	exper	1.085766	.0221187	4.04	0.000	1.043268	1.129995
pyperstudlep		1.032768	.0196238	1.70	0.090	.9950131	1.071955
pyperstude~s		1.030538	.0077901	3.98	0.000	1.015383	1.04592
pyperstudw~e		.9700115	.048562	-0.61	0.543	.8793523	1.070017
pyperstudb~k		.9640894	.0496442	-0.71	0.478	.8715375	1.06647
pyperstudl~o		.9442364	.0471441	-1.15	0.250	.8562127	1.041309
pyperstudm~e		1.009205	.0084178	1.10	0.272	.99284	1.025839
	male	1.784462	.3695297	2.80	0.005	1.189156	2.677785
	age	.9516836	.0177883	-2.65	0.008	.9174501	.9871945
	Asian	3.52e-06	.0029681	-0.01	0.988	0	.
	Black	2.147991	1.220848	1.35	0.179	.705078	6.543765
	Latino	.5222456	.4536267	-0.75	0.455	.0951731	2.865729
	Other	1.159718	1.833534	0.09	0.925	.0523104	25.71087
	AsianMale	(omitted)					
	BlackMale	.4471096	.3803837	-0.95	0.344	.084381	2.369101
	LatinoMale	9.876086	14.39639	1.57	0.116	.5672656	171.9425
	OtherMale	1.12e-06	.0009437	-0.02	0.987	0	.

-----> region = 7

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -1013.0522
Iteration 1: log likelihood = -952.41468
Iteration 2: log likelihood = -950.05734
Iteration 3: log likelihood = -949.89257
Iteration 4: log likelihood = -949.86076

```

```

Iteration 5: log likelihood = -949.85396
Iteration 6: log likelihood = -949.85239
Iteration 7: log likelihood = -949.85202
Iteration 8: log likelihood = -949.85194
Iteration 9: log likelihood = -949.85193

```

```

Multinomial logistic regression
Number of obs   =      1006
LR chi2(34)     =      126.40
Prob > chi2     =      0.0000
Pseudo R2      =      0.0624

Log likelihood = -949.85193

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.002986	.0132394	0.23	0.821	.9773704	1.029274
exper	.995901	.0175231	-0.23	0.815	.9621419	1.030845
pypersistudlep	1.006757	.0221476	0.31	0.759	.9642715	1.051115
pypersistude~s	.9935448	.0076473	-0.84	0.400	.9786689	1.008647
pypersistudw~e	.8319614	.0829198	-1.85	0.065	.6843302	1.011441
pypersistudb~k	.8368125	.0845892	-1.76	0.078	.6864114	1.020168
pypersistudl~o	.8357957	.0844817	-1.77	0.076	.6855848	1.018918
pypersistudm~e	.9789413	.0121372	-1.72	0.086	.9554396	1.003021
male	1.959335	.4168825	3.16	0.002	1.291222	2.973149
age	.966271	.0154443	-2.15	0.032	.93647	.9970205
Asian	6.88e-07	.0012563	-0.01	0.994	0	.
Black	2.726635	.9242021	2.96	0.003	1.403167	5.298398
Latino	.4312152	.4800978	-0.76	0.450	.0486407	3.822857
Other	9.27e-07	.0011474	-0.01	0.991	0	.
AsianMale	(omitted)					
BlackMale	1.06512	.5096754	0.13	0.895	.4169523	2.720889
LatinoMale	1.531478	2.125252	0.31	0.759	.1008992	23.24525
OtherMale	.7616127	1323.623	-0.00	1.000	0	.
3						
tenure	.9803089	.0096853	-2.01	0.044	.9615087	.9994767
exper	1.036938	.0146051	2.58	0.010	1.008704	1.065962
pypersistudlep	1.008662	.0182238	0.48	0.633	.973569	1.04502
pypersistude~s	1.030698	.0060774	5.13	0.000	1.018855	1.042679
pypersistudw~e	.8278154	.0738847	-2.12	0.034	.6949627	.9860649
pypersistudb~k	.8084624	.0731232	-2.35	0.019	.6771283	.9652697
pypersistudl~o	.8092009	.0730747	-2.34	0.019	.6779362	.9658815
pypersistudm~e	1.013646	.0066774	2.06	0.040	1.000643	1.026818
male	1.676492	.2787148	3.11	0.002	1.210293	2.32227
age	.9870027	.0129219	-1.00	0.318	.9619985	1.012657
Asian	1.80252	2.657486	0.40	0.689	.1002164	32.42062
Black	1.513949	.4794361	1.31	0.190	.8138667	2.816236
Latino	.2742771	.2987966	-1.19	0.235	.032426	2.319989
Other	.8045956	.9453392	-0.19	0.853	.0804415	8.047766
AsianMale	(omitted)					
BlackMale	1.127382	.5271245	0.26	0.798	.4508984	2.818793
LatinoMale	4.751992	6.090403	1.22	0.224	.385417	58.5896
OtherMale	1.68847	2.653133	0.33	0.739	.0776211	36.72881

```
-----> region = 8
```

```

note: AsianMale omitted because of collinearity
note: LatinoMale omitted because of collinearity
Iteration 0: log likelihood = -339.81399
Iteration 1: log likelihood = -304.47623
Iteration 2: log likelihood = -302.40627
Iteration 3: log likelihood = -302.21687
Iteration 4: log likelihood = -302.17674

```



```

Iteration 5:  log likelihood = -302.16775
Iteration 6:  log likelihood = -302.1656
Iteration 7:  log likelihood = -302.16514
Iteration 8:  log likelihood = -302.16504
Iteration 9:  log likelihood = -302.16502

```

```

Multinomial logistic regression      Number of obs   =      337
                                     LR chi2(32)        =      75.30
                                     Prob > chi2         =      0.0000
Log likelihood = -302.16502          Pseudo R2       =      0.1108

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2	tenure	.9831057	.0239139	-0.70	0.484	.937335	1.031111
	exper	1.030464	.0319304	0.97	0.333	.9697443	1.094986
pypersistudlep		1.066613	.0657452	1.05	0.295	.945234	1.203578
pypersistude~s		.9767726	.0160717	-1.43	0.153	.9457752	1.008786
pypersistudw~e		.7671826	.1949819	-1.04	0.297	.4661902	1.262509
pypersistudb~k		.8077241	.2020946	-0.85	0.393	.4946412	1.318973
pypersistudl~o		.7392726	.186575	-1.20	0.231	.4507974	1.21235
pypersistudm~e		.9575594	.0315914	-1.31	0.189	.8976008	1.021523
	male	1.496195	.5709565	1.06	0.291	.7082146	3.160906
	age	.9817791	.0282313	-0.64	0.522	.9279771	1.0387
	Asian	5.11e+07	2.07e+11	0.00	0.996	0	.
	Black	1.643719	1.306586	0.63	0.532	.3461071	7.806288
	Latino	1.113687	1.480266	0.08	0.935	.0822966	15.07109
	Other	1.27e-06	.0010721	-0.02	0.987	0	.
	AsianMale	(omitted)					
	BlackMale	1.526528	1.647963	0.39	0.695	.1839885	12.6654
	LatinoMale	(omitted)					
	OtherMale	8.66e+11	1.02e+15	0.02	0.981	0	.
3	tenure	.9998667	.0180679	-0.01	0.994	.9650741	1.035914
	exper	1.060122	.0276966	2.23	0.025	1.007204	1.11582
pypersistudlep		1.057461	.0426581	1.39	0.166	.977073	1.144464
pypersistude~s		1.034302	.011412	3.06	0.002	1.012175	1.056913
pypersistudw~e		1.184993	.2162887	0.93	0.352	.8286129	1.694649
pypersistudb~k		1.140619	.2054597	0.73	0.465	.8013338	1.623557
pypersistudl~o		1.098776	.2012603	0.51	0.607	.7673556	1.573338
pypersistudm~e		1.007371	.0134818	0.55	0.583	.9812908	1.034145
	male	2.119846	.6003027	2.65	0.008	1.216912	3.692747
	age	.9809577	.0239053	-0.79	0.430	.9352056	1.028948
	Asian	.7629208	4693.692	-0.00	1.000	0	.
	Black	2.528545	1.55559	1.51	0.132	.7571844	8.443834
	Latino	1.182312	1.195422	0.17	0.868	.1629644	8.577716
	Other	2.83e-06	.001612	-0.02	0.982	0	.
	AsianMale	(omitted)					
	BlackMale	.5152623	.5022497	-0.68	0.496	.0762647	3.481236
	LatinoMale	(omitted)					
	OtherMale	2.13e+11	2.12e+14	0.03	0.979	0	.

```

-----> region = 9

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0:  log likelihood = -220.98733
Iteration 1:  log likelihood = -193.46518
Iteration 2:  log likelihood = -191.37818
Iteration 3:  log likelihood = -191.03453

```

```

Iteration 4: log likelihood = -190.96668
Iteration 5: log likelihood = -190.94951
Iteration 6: log likelihood = -190.94614
Iteration 7: log likelihood = -190.94563
Iteration 8: log likelihood = -190.94557
Iteration 9: log likelihood = -190.94556

```

```

Multinomial logistic regression
Number of obs   =      209
LR chi2(30)     =      60.08
Prob > chi2     =      0.0009
Pseudo R2      =      0.1359

Log likelihood = -190.94556

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	tenure	1.000994	.0217372	0.05	0.964	.9592835	1.044517
	exper	1.021906	.0388248	0.57	0.568	.9485747	1.100906
	pyperstudlep	1.065684	.0610401	1.11	0.267	.9525185	1.192294
	pyperstude~s	.9503489	.0143102	-3.38	0.001	.9227112	.9788143
	pyperstudw~e	.887498	.1165541	-0.91	0.363	.6860884	1.148034
	pyperstudb~k	.9001962	.1330886	-0.71	0.477	.6737396	1.202769
	pyperstudl~o	.9178581	.1187574	-0.66	0.508	.7122665	1.182792
	pyperstudm~e	.9857519	.0189719	-0.75	0.456	.9492604	1.023646
	male	.5334651	.1886309	-1.78	0.076	.2667626	1.06681
	age	.9784408	.0328694	-0.65	0.516	.916093	1.045032
	Asian	(omitted)					
	Black	.2595343	.3521123	-0.99	0.320	.0181703	3.707037
	Latino	.965088.9	1.39e+09	0.01	0.992	0	.
	Other	4.34e-07	.0008028	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.7344552	6.20e+09	0.02	0.985	0	.
	LatinoMale	4.45e-13	9.77e-10	-0.01	0.990	0	.
	OtherMale	(omitted)					
2							
	tenure	1.033173	.0349898	0.96	0.335	.9668212	1.104079
	exper	.9024222	.0420719	-2.20	0.028	.823618	.9887664
	pyperstudlep	1.110988	.0815819	1.43	0.152	.9620642	1.282965
	pyperstude~s	.9486948	.0193351	-2.58	0.010	.9115457	.9873579
	pyperstudw~e	.8268603	.1261518	-1.25	0.213	.6131499	1.115059
	pyperstudb~k	.8733181	.1480339	-0.80	0.424	.6264515	1.217468
	pyperstudl~o	.8315379	.1267838	-1.21	0.226	.6167372	1.121151
	pyperstudm~e	1.013907	.0205459	0.68	0.496	.9744275	1.054987
	male	.9727765	.4512665	-0.06	0.953	.3918724	2.414802
	age	1.011987	.036058	0.33	0.738	.9437258	1.085185
	Asian	(omitted)					
	Black	1.465647	1.620081	0.35	0.729	.1679337	12.79149
	Latino	.1255378	1.81e+09	0.01	0.992	0	.
	Other	3.955612	5.684369	0.96	0.339	.236594	66.13381
	AsianMale	(omitted)					
	BlackMale	.622578.6	5.25e+08	0.02	0.987	0	.
	LatinoMale	3.83e-13	1.04e-09	-0.01	0.992	0	.
	OtherMale	(omitted)					
3		(base outcome)					

```

-----> region = 10

```

```

Iteration 0: log likelihood = -3218.098
Iteration 1: log likelihood = -3044.4752
Iteration 2: log likelihood = -3042.3541
Iteration 3: log likelihood = -3042.3511
Iteration 4: log likelihood = -3042.3511

```

Multinomial logistic regression

Number of obs = 2998

LR chi2(36) = 351.49

Prob > chi2 = 0.0000

Pseudo R2 = 0.0546

Log likelihood = -3042.3511

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.016467	.0069489	2.39	0.017	1.002938	1.030178
exper	.9852188	.0091739	-1.60	0.110	.9674014	1.003364
pyperstudlep	1.000954	.0055988	0.17	0.865	.9900401	1.011987
pyperstude~s	.9901731	.0035368	-2.76	0.006	.9832652	.9971295
pyperstudw~e	.985211	.0093259	-1.57	0.115	.9671011	1.00366
pyperstudb~k	.9909935	.0097288	-0.92	0.357	.9721077	1.010246
pyperstudl~o	.9927181	.0095552	-0.76	0.448	.9741658	1.011624
pyperstudm~e	.9980477	.0048201	-0.40	0.686	.988645	1.00754
male	1.410692	.1580124	3.07	0.002	1.132631	1.757016
age	.9677835	.0076344	-4.15	0.000	.9529355	.9828628
Asian	.8524511	.6221997	-0.22	0.827	.2038834	3.56416
Black	1.548927	.2490868	2.72	0.007	1.130179	2.122828
Latino	2.614627	.5831117	4.31	0.000	1.688789	4.048031
Other	.7140717	.4197304	-0.57	0.567	.2256351	2.259836
AsianMale	1.870162	1.844786	0.63	0.526	.2705375	12.92799
BlackMale	1.730054	.4249049	2.23	0.026	1.069062	2.799733
LatinoMale	.8721867	.2980691	-0.40	0.689	.4463875	1.704146
OtherMale	1.055083	.8206907	0.07	0.945	.2297119	4.846066
3						
tenure	1.025446	.0065099	3.96	0.000	1.012766	1.038285
exper	1.04074	.0103783	4.00	0.000	1.020596	1.061281
pyperstudlep	1.017565	.0063194	2.80	0.005	1.005254	1.030026
pyperstude~s	1.007315	.0035857	2.05	0.041	1.000311	1.014367
pyperstudw~e	1.021306	.0102258	2.11	0.035	1.001459	1.041546
pyperstudb~k	1.003011	.0105925	0.28	0.776	.9824634	1.023988
pyperstudl~o	.9904894	.0101918	-0.93	0.353	.970714	1.010668
pyperstudm~e	1.010858	.0046857	2.33	0.020	1.001716	1.020083
male	1.206042	.1360975	1.66	0.097	.9667345	1.504587
age	.9775841	.0087109	-2.54	0.011	.9606591	.9948072
Asian	.8127482	.6037372	-0.28	0.780	.1895197	3.48544
Black	1.47786	.2572572	2.24	0.025	1.050658	2.078766
Latino	2.601955	.6081815	4.09	0.000	1.645659	4.113957
Other	.3378231	.2631703	-1.39	0.164	.0733818	1.555216
AsianMale	1.977114	2.028312	0.66	0.506	.2647207	14.76643
BlackMale	1.462103	.3998428	1.39	0.165	.8554585	2.498946
LatinoMale	1.327486	.4789514	0.79	0.432	.6545152	2.692402
OtherMale	1.485554	1.512756	0.39	0.698	.2018834	10.93142

-----> region = 11

Iteration 0: log likelihood = -2299.5528  
Iteration 1: log likelihood = -2152.3809  
Iteration 2: log likelihood = -2149.8626  
Iteration 3: log likelihood = -2149.7495  
Iteration 4: log likelihood = -2149.7344  
Iteration 5: log likelihood = -2149.7325  
Iteration 6: log likelihood = -2149.7322  
Iteration 7: log likelihood = -2149.7321  
Iteration 8: log likelihood = -2149.7321

Multinomial logistic regression

Number of obs = 2112

LR chi2(36) = 299.64

Prob > chi2 = 0.0000

Pseudo R2 = 0.0652

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.003514	.0080841	0.44	0.663	.9877935	1.019484
	exper	.9953014	.0099964	-0.47	0.639	.9759003	1.015088
pyp	erstudlep	.9888107	.0078549	-1.42	0.157	.9735346	1.004326
pyp	erstud~s	.9962936	.0039709	-0.93	0.352	.988541	1.004107
pyp	erstudw~e	.9845133	.0145428	-1.06	0.291	.9564186	1.013433
pyp	erstudb~k	.988565	.01529	-0.74	0.457	.9590468	1.018992
pyp	erstudl~o	.9977121	.0150226	-0.15	0.879	.9686985	1.027595
pyp	erstudm~e	.989335	.0043524	-2.44	0.015	.9808411	.9979024
	male	1.046136	.125773	0.38	0.708	.8265162	1.324113
	age	.9697608	.0085338	-3.49	0.000	.9531783	.9866319
	Asian	1.060571	.820199	0.08	0.939	.2329474	4.828604
	Black	2.156	.5254969	3.15	0.002	1.337148	3.476306
	Latino	2.89278	.9278709	3.31	0.001	1.542727	5.424275
	Other	1.076982	.6650058	0.12	0.904	.3210814	3.612445
	AsianMale	.9911519	1.260614	-0.01	0.994	.0819455	11.98824
	BlackMale	2.133478	.8542911	1.89	0.058	.9733042	4.676574
	LatinoMale	.7546765	.3603227	-0.59	0.556	.296041	1.923844
	OtherMale	2.162942	2.41645	0.69	0.490	.2421466	19.3202
3							
	tenure	1.006951	.0076024	0.92	0.359	.99216	1.021962
	exper	1.067085	.0117641	5.89	0.000	1.044275	1.090394
pyp	erstudlep	1.004308	.0080807	0.53	0.593	.9885942	1.020271
pyp	erstud~s	1.025009	.0041562	6.09	0.000	1.016896	1.033188
pyp	erstudw~e	1.039534	.0173432	2.32	0.020	1.006091	1.074088
pyp	erstudb~k	1.018487	.0179093	1.04	0.298	.9839834	1.0542
pyp	erstudl~o	1.013996	.0174216	0.81	0.419	.9804185	1.048723
pyp	erstudm~e	1.001214	.004046	0.30	0.764	.9933152	1.009175
	male	1.208504	.1497281	1.53	0.126	.947956	1.540665
	age	.9768267	.0098749	-2.32	0.020	.9576626	.9963742
	Asian	.5045433	.5686115	-0.61	0.544	.0554127	4.593965
	Black	1.52234	.442931	1.44	0.149	.8606988	2.692602
	Latino	2.817875	.9559118	3.05	0.002	1.44933	5.478682
	Other	.5963186	.4968377	-0.62	0.535	.1164873	3.052657
	AsianMale	.0000108	.0060308	-0.02	0.984	0	.
	BlackMale	1.890253	.8811625	1.37	0.172	.7580935	4.713213
	LatinoMale	.5087433	.2788192	-1.23	0.218	.1737776	1.489374
	OtherMale	5.156694	6.461226	1.31	0.190	.4424194	60.10472

```
-----> region = 12
```

```
note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -838.86662
Iteration 1: log likelihood = -777.89605
Iteration 2: log likelihood = -775.70764
Iteration 3: log likelihood = -775.54051
Iteration 4: log likelihood = -775.50372
Iteration 5: log likelihood = -775.4962
Iteration 6: log likelihood = -775.49499
Iteration 7: log likelihood = -775.49469
Iteration 8: log likelihood = -775.49463
Iteration 9: log likelihood = -775.49462
```

Multinomial logistic regression	Number of obs	=	788
	LR chi2(32)	=	126.74
	Prob > chi2	=	0.0000

Log likelihood = -775.49462

Pseudo R2 = 0.0755

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.002086	.0136498	0.15	0.878	.9756868	1.029199
	exper	.9775908	.0176514	-1.26	0.209	.9435996	1.012806
	pyperstudlep	1.04704	.0249877	1.93	0.054	.9991932	1.097179
	pyperstude~s	.9898814	.0070525	-1.43	0.153	.9761549	1.003801
	pyperstudw~e	.8864232	.0475481	-2.25	0.025	.7979622	.9846909
	pyperstudb~k	.9073811	.0520749	-1.69	0.090	.8108472	1.015408
	pyperstudl~o	.8922187	.0484798	-2.10	0.036	.8020847	.9924816
	pyperstudm~e	.9938118	.0078938	-0.78	0.435	.9784602	1.009404
	male	1.140835	.2506693	0.60	0.549	.7416401	1.754901
	age	.9661091	.0155944	-2.14	0.033	.936023	.9971622
	Asian	(omitted)					
	Black	.8231653	.2949363	-0.54	0.587	.4078548	1.661379
	Latino	1.838054	1.156445	0.97	0.333	.5355617	6.308225
	Other	3.41e-07	.0003562	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	2.02144	1.239933	1.15	0.251	.6074935	6.726358
	LatinoMale	1.646068	1.553117	0.53	0.597	.2590073	10.46125
	OtherMale	1.76788	2970.986	0.00	1.000	0	.
3							
	tenure	1.013111	.0110361	1.20	0.232	.9917095	1.034974
	exper	1.035038	.0172298	2.07	0.039	1.001813	1.069365
	pyperstudlep	1.057896	.0229136	2.60	0.009	1.013926	1.103773
	pyperstude~s	1.011552	.0062388	1.86	0.063	.9993974	1.023854
	pyperstudw~e	.9999737	.0548549	-0.00	1.000	.8980381	1.11348
	pyperstudb~k	.9879766	.0574504	-0.21	0.835	.8815554	1.107245
	pyperstudl~o	.9720559	.053788	-0.51	0.609	.872149	1.083407
	pyperstudm~e	1.007253	.0067676	1.08	0.282	.9940759	1.020605
	male	1.358045	.255241	1.63	0.103	.9395786	1.962887
	age	.9767109	.0148533	-1.55	0.121	.9480285	1.006261
	Asian	(omitted)					
	Black	.8030415	.2957905	-0.60	0.552	.3901291	1.65298
	Latino	.8073375	.6236338	-0.28	0.782	.177639	3.669205
	Other	6.90e-07	.0006594	-0.01	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	1.732019	1.053941	0.90	0.367	.5255249	5.708371
	LatinoMale	1.337187	1.612281	0.24	0.810	.1258567	14.20718
	OtherMale	.7482322	1127.163	-0.00	1.000	0	.

-----> region = 13

Iteration 0: log likelihood = -1374.1326  
 Iteration 1: log likelihood = -1291.899  
 Iteration 2: log likelihood = -1290.0895  
 Iteration 3: log likelihood = -1289.8589  
 Iteration 4: log likelihood = -1289.8063  
 Iteration 5: log likelihood = -1289.7958  
 Iteration 6: log likelihood = -1289.7941  
 Iteration 7: log likelihood = -1289.7937  
 Iteration 8: log likelihood = -1289.7936  
 Iteration 9: log likelihood = -1289.7935

Multinomial logistic regression

Number of obs = 1258  
 LR chi2(36) = 168.68  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0614

Log likelihood = -1289.7935

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.026672	.011443	2.36	0.018	1.004487	1.049346
	exper	.9518725	.012361	-3.80	0.000	.927951	.9764107
pyperstudlep		.9992728	.0112176	-0.06	0.948	.9775267	1.021503
pyperstude~s		.9851096	.0071634	-2.06	0.039	.9711692	.99925
pyperstudw~e		.9838681	.0240724	-0.66	0.506	.9378006	1.032199
pyperstudb~k		1.000637	.0275113	0.02	0.982	.9481432	1.056038
pyperstudl~o		.9994484	.024625	-0.02	0.982	.9523311	1.048897
pyperstudm~e		.9903055	.0060668	-1.59	0.112	.9784858	1.002268
	male	.7871632	.1340598	-1.41	0.160	.5637659	1.099084
	age	.9873338	.0116661	-1.08	0.281	.9647315	1.010466
	Asian	.6014127	.7500563	-0.41	0.683	.0521904	6.930345
	Black	1.920773	.643344	1.95	0.051	.9962645	3.703204
	Latino	1.922135	.5015123	2.50	0.012	1.152639	3.205343
	Other	4.079122	4.970755	1.15	0.249	.3743739	44.44549
	AsianMale	2.71e+07	1.11e+11	0.00	0.997	0	.
	BlackMale	2.076445	1.185079	1.28	0.200	.6784508	6.355101
	LatinoMale	2.358145	.9973417	2.03	0.043	1.029361	5.402233
	OtherMale	.3446166	.5107409	-0.72	0.472	.0188716	6.293095
3							
	tenure	1.009994	.0104092	0.96	0.335	.989797	1.030603
	exper	1.014311	.0134787	1.07	0.285	.9882347	1.041076
pyperstudlep		.9899915	.0109694	-0.91	0.364	.9687236	1.011726
pyperstude~s		1.025177	.0070461	3.62	0.000	1.011459	1.03908
pyperstudw~e		.9794749	.0249702	-0.81	0.416	.9317369	1.029659
pyperstudb~k		.9605449	.0276594	-1.40	0.162	.9078349	1.016315
pyperstudl~o		.9703164	.0251066	-1.16	0.244	.9223353	1.020793
pyperstudm~e		.9956455	.0058099	-0.75	0.455	.9843232	1.007098
	male	1.138143	.1815498	0.81	0.417	.8325651	1.555877
	age	.9851425	.0124446	-1.18	0.236	.961051	1.009838
	Asian	1.157422	1.43941	0.12	0.906	.1011367	13.2457
	Black	.77019	.3203868	-0.63	0.530	.3408094	1.740541
	Latino	.9853375	.2883596	-0.05	0.960	.5552395	1.748597
	Other	6.06e-07	.0009011	-0.01	0.992	0	.
	AsianMale	1.322026	8274.767	0.00	1.000	0	.
	BlackMale	2.74761	1.782102	1.56	0.119	.7706643	9.795913
	LatinoMale	2.102352	.963106	1.62	0.105	.856569	5.159984
	OtherMale	.8483886	1587.648	-0.00	1.000	0	.

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -282.75147  
Iteration 1: log likelihood = -249.10448  
Iteration 2: log likelihood = -245.72455  
Iteration 3: log likelihood = -244.67042  
Iteration 4: log likelihood = -244.56038  
Iteration 5: log likelihood = -244.54056  
Iteration 6: log likelihood = -244.53731  
Iteration 7: log likelihood = -244.53653  
Iteration 8: log likelihood = -244.53636  
Iteration 9: log likelihood = -244.53632  
Iteration 10: log likelihood = -244.53632

Multinomial logistic regression

Number of obs = 298  
LR chi2(30) = 76.43  
Prob > chi2 = 0.0000

Log likelihood = -244.53632

Pseudo R2 = 0.1352

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.014839	.0332925	0.45	0.653	.9516404	1.082235
	exper	.9559811	.0407086	-1.06	0.290	.8794326	1.039193
	pyperstudlep	.9485245	.1121891	-0.45	0.655	.7522644	1.195987
	pyperstude~s	.9802241	.0201874	-0.97	0.332	.9414454	1.0206
	pyperstudw~e	.8974237	.2494892	-0.39	0.697	.5204252	1.547522
	pyperstudb~k	.9256226	.273646	-0.26	0.794	.5185485	1.65226
	pyperstudl~o	.9073812	.2485292	-0.35	0.723	.5304544	1.552142
	pyperstudm~e	1.01722	.0267611	0.65	0.516	.9660985	1.071047
	male	2.165113	.9952842	1.68	0.093	.8794081	5.330534
	age	1.023775	.0394241	0.61	0.542	.9493488	1.104035
	Asian	(omitted)					
	Black	1.87e-06	.002352	-0.01	0.992	0	.
	Latino	1.10e-06	.0011184	-0.01	0.989	0	.
	Other	5.33e-07	.0017153	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	1.96e+13	5.69e+16	0.01	0.992	0	.
	LatinoMale	2007125	2.03e+09	0.01	0.989	0	.
	OtherMale	(omitted)					
3							
	tenure	1.003074	.0197742	0.16	0.876	.965056	1.042589
	exper	1.076946	.0315305	2.53	0.011	1.016887	1.140552
	pyperstudlep	1.099862	.052381	2.00	0.046	1.001843	1.207471
	pyperstude~s	1.034374	.0131693	2.65	0.008	1.008882	1.06051
	pyperstudw~e	1.365689	.2502485	1.70	0.089	.9536256	1.955806
	pyperstudb~k	1.325987	.2580808	1.45	0.147	.9054555	1.941829
	pyperstudl~o	1.317514	.239871	1.51	0.130	.9221099	1.882469
	pyperstudm~e	1.009424	.0178375	0.53	0.596	.9750618	1.044997
	male	2.856741	.8455011	3.55	0.000	1.599349	5.102682
	age	.9370753	.0257122	-2.37	0.018	.8880114	.98885
	Asian	(omitted)					
	Black	3.542016	3.861233	1.16	0.246	.4181539	30.00302
	Latino	.2164203	.2638596	-1.26	0.209	.0198387	2.360928
	Other	1.48e-07	.0002445	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	3800365	9.96e+09	0.01	0.995	0	.
	LatinoMale	2.629369	4.028998	0.63	0.528	.1304819	52.98498
	OtherMale	(omitted)					

-----> region = 15

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -321.49229  
Iteration 1: log likelihood = -289.41107  
Iteration 2: log likelihood = -286.80617  
Iteration 3: log likelihood = -286.6592  
Iteration 4: log likelihood = -286.62585  
Iteration 5: log likelihood = -286.61929  
Iteration 6: log likelihood = -286.61821  
Iteration 7: log likelihood = -286.61795  
Iteration 8: log likelihood = -286.6179  
Iteration 9: log likelihood = -286.61789

Multinomial logistic regression

Number of obs = 317  
LR chi2(30) = 69.75

Log likelihood = -286.61789      Prob > chi2 = 0.0001  
Pseudo R2 = 0.1085

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.024901	.02992	0.84	0.399	.9679052	1.085254
exper	.9828828	.0367415	-0.46	0.644	.9134456	1.057598
pyperstudlep	1.021156	.0247979	0.86	0.389	.973692	1.070935
pyperstude~s	.9844654	.0128727	-1.20	0.231	.959556	1.010021
pyperstudw~e	.6389755	.1742464	-1.64	0.100	.3744246	1.090445
pyperstudb~k	.6858027	.200122	-1.29	0.196	.3870907	1.215026
pyperstudl~o	.6427249	.1746364	-1.63	0.104	.3773488	1.094731
pyperstudm~e	1.021553	.0141524	1.54	0.124	.9941885	1.049672
male	2.433286	1.102055	1.96	0.050	1.001561	5.911657
age	.9793889	.0328233	-0.62	0.534	.9171237	1.045881
Asian	(omitted)					
Black	2.19e-06	.0045477	-0.01	0.995	0	.
Latino	7.977139	4.96656	3.34	0.001	2.354451	27.02742
Other	.3687587	1065.428	-0.00	1.000	0	.
AsianMale	(omitted)					
BlackMale	.3085392	779.572	-0.00	1.000	0	.
LatinoMale	.3016061	.2361383	-1.53	0.126	.0650136	1.399189
OtherMale	(omitted)					
3						
tenure	.986858	.0188037	-0.69	0.487	.9506832	1.024409
exper	1.063398	.0307272	2.13	0.033	1.004847	1.12536
pyperstudlep	1.020935	.0215017	0.98	0.325	.9796506	1.06396
pyperstude~s	1.028397	.0102728	2.80	0.005	1.008458	1.04873
pyperstudw~e	1.13799	.2392517	0.61	0.539	.7536702	1.718287
pyperstudb~k	1.156518	.2606318	0.65	0.519	.743578	1.798781
pyperstudl~o	1.111914	.2336161	0.50	0.614	.7365999	1.67846
pyperstudm~e	.9965469	.0133996	-0.26	0.797	.9706271	1.023159
male	1.653308	.5040672	1.65	0.099	.9095698	3.005187
age	.9583854	.0252766	-1.61	0.107	.9101028	1.009229
Asian	(omitted)					
Black	3.71e-06	.005165	-0.01	0.993	0	.
Latino	.6800798	.4588023	-0.57	0.568	.1812649	2.551561
Other	.5878507	8.51e+09	0.01	0.991	0	.
AsianMale	(omitted)					
BlackMale	.2355186	401.2442	-0.00	0.999	0	.
LatinoMale	1.401016	1.137914	0.42	0.678	.2851622	6.883262
OtherMale	(omitted)					

-----> region = 16

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -450.0077  
Iteration 1: log likelihood = -394.56534  
Iteration 2: log likelihood = -390.49532  
Iteration 3: log likelihood = -390.17668  
Iteration 4: log likelihood = -390.10564  
Iteration 5: log likelihood = -390.09065  
Iteration 6: log likelihood = -390.08756  
Iteration 7: log likelihood = -390.08686  
Iteration 8: log likelihood = -390.08669  
Iteration 9: log likelihood = -390.08665  
Iteration 10: log likelihood = -390.08664

Multinomial logistic regression

Number of obs = 432  
LR chi2 (34) = 119.84



Log likelihood = -390.08664

Prob > chi2 = 0.0000  
Pseudo R2 = 0.1332

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.056582	.0266933	2.18	0.029	1.005538	1.110217
	exper	.9225433	.0303142	-2.45	0.014	.8650014	.983913
pyperstudlep		.9773562	.0285801	-0.78	0.433	.9229152	1.035009
pyperstude~s		.9818035	.0105445	-1.71	0.087	.9613526	1.002689
pyperstudw~e		.9096858	.0345078	-2.50	0.013	.8445048	.9798977
pyperstudb~k		.9549968	.0431242	-1.02	0.308	.8741073	1.043372
pyperstudl~o		.9228832	.034385	-2.15	0.031	.8578918	.9927982
pyperstudm~e		1.032493	.0154947	2.13	0.033	1.002567	1.063314
	male	1.66677	.545033	1.56	0.118	.8780783	3.163867
	age	.9848093	.0274499	-0.55	0.583	.9324517	1.040107
	Asian	3.82e+07	2.33e+11	0.00	0.998	0	.
	Black	.375252	.5000833	-0.74	0.462	.0275398	5.113112
	Latino	4.092647	2.651437	2.18	0.030	1.149609	14.56996
	Other	.590198	.8191962	-0.38	0.704	.0388623	8.963289
	AsianMale	(omitted)					
	BlackMale	1.57e-07	.0001855	-0.01	0.989	0	.
	LatinoMale	.380751	.4314056	-0.85	0.394	.0413235	3.508207
	OtherMale	2.41e+07	3.93e+10	0.01	0.992	0	.
3							
	tenure	1.04627	.0168855	2.80	0.005	1.013693	1.079894
	exper	.9872598	.0235071	-0.54	0.590	.9422452	1.034425
pyperstudlep		1.07342	.0206168	3.69	0.000	1.033763	1.114599
pyperstude~s		1.005959	.0080654	0.74	0.459	.990275	1.021892
pyperstudw~e		1.039743	.0406432	1.00	0.319	.9630586	1.122533
pyperstudb~k		1.028836	.0453515	0.64	0.519	.9436804	1.121676
pyperstudl~o		1.012058	.038989	0.31	0.756	.9384549	1.091434
pyperstudm~e		1.01435	.01388	1.04	0.298	.9875072	1.041922
	male	1.830166	.4566007	2.42	0.015	1.122348	2.984374
	age	1.020616	.0229005	0.91	0.363	.9767044	1.066502
	Asian	1.653238	15218.19	0.00	1.000	0	.
	Black	.5576694	.6854873	-0.48	0.635	.0501279	6.204029
	Latino	.3750554	.3878043	-0.95	0.343	.0494265	2.845975
	Other	5.84e-07	.0006487	-0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	.7379837	1.332548	-0.17	0.866	.0214321	25.41136
	LatinoMale	2.695601	3.63901	0.73	0.463	.1912272	37.99806
	OtherMale	1.62e+12	3.19e+15	0.01	0.989	0	.

-----> region = 17

note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -463.36597  
Iteration 1: log likelihood = -407.11943  
Iteration 2: log likelihood = -405.62968  
Iteration 3: log likelihood = -405.57229  
Iteration 4: log likelihood = -405.55977  
Iteration 5: log likelihood = -405.55716  
Iteration 6: log likelihood = -405.55673  
Iteration 7: log likelihood = -405.55663  
Iteration 8: log likelihood = -405.5566

Multinomial logistic regression

Number of obs = 464  
LR chi2(34) = 115.62  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.1248

Log likelihood = -405.5566

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.009415	.0211253	0.45	0.654	.9688475	1.05168
exper	1.01802	.0322354	0.56	0.573	.9567607	1.083202
pyperstudlep	1.025921	.0477758	0.55	0.583	.9364284	1.123966
pyperstude~s	.9764718	.0100641	-2.31	0.021	.9569445	.9963976
pyperstudw~e	.997722	.1526101	-0.01	0.988	.7392836	1.346505
pyperstudb~k	1.028829	.1560467	0.19	0.851	.7642545	1.384996
pyperstudl~o	1.013978	.1529348	0.09	0.927	.7544745	1.362737
pyperstudm~e	.9855151	.0106389	-1.35	0.177	.9648823	1.006589
male	2.365627	.8025416	2.54	0.011	1.216677	4.599571
age	1.010118	.0290303	0.35	0.726	.9547928	1.06865
Asian	4.83e-06	.0064444	-0.01	0.993	0	.
Black	1.948853	1.90714	0.68	0.495	.2862836	13.26666
Latino	1.12041	1.255481	0.10	0.919	.1246118	10.07384
Other	2.835766	4.109845	0.72	0.472	.1655895	48.56326
AsianMale	.5848826	1104.63	-0.00	1.000	0	.
BlackMale	2.168524	3.363033	0.50	0.618	.1037736	45.31496
LatinoMale	2.284791	3.094016	0.61	0.542	.1607561	32.47324
OtherMale	(omitted)					
3						
tenure	1.008954	.0155689	0.58	0.563	.9788967	1.039935
exper	1.028708	.0243215	1.20	0.231	.9821268	1.077499
pyperstudlep	1.093904	.0326109	3.01	0.003	1.031819	1.159724
pyperstude~s	1.013909	.0095615	1.46	0.143	.9953409	1.032824
pyperstudw~e	1.528758	.2127941	3.05	0.002	1.163741	2.008264
pyperstudb~k	1.500959	.2088228	2.92	0.004	1.142732	1.971485
pyperstudl~o	1.500549	.2077494	2.93	0.003	1.143937	1.968331
pyperstudm~e	.9891958	.0087013	-1.23	0.217	.9722877	1.006398
male	3.279379	.8209268	4.74	0.000	2.007754	5.356399
age	1.007513	.0222369	0.34	0.735	.9648588	1.052053
Asian	4.42e-06	.0033642	-0.02	0.987	0	.
Black	1.158699	1.106215	0.15	0.877	.1783711	7.526906
Latino	1.846647	1.131477	1.00	0.317	.5556944	6.136657
Other	1.345015	1.717097	0.23	0.816	.1101669	16.42114
AsianMale	.5722363	616.0003	-0.00	1.000	0	.
BlackMale	2.192966	3.383227	0.51	0.611	.1066177	45.10601
LatinoMale	1.598777	1.399878	0.54	0.592	.2873965	8.893942
OtherMale	(omitted)					

-----> region = 18

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -399.36483  
Iteration 1: log likelihood = -360.72429  
Iteration 2: log likelihood = -358.90308  
Iteration 3: log likelihood = -358.79246  
Iteration 4: log likelihood = -358.76962  
Iteration 5: log likelihood = -358.76444  
Iteration 6: log likelihood = -358.76355  
Iteration 7: log likelihood = -358.76345  
Iteration 8: log likelihood = -358.76343

Multinomial logistic regression

Number of obs = 392  
LR chi2(32) = 81.20  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.1017

Log likelihood = -358.76343

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.014155	.0237158	0.60	0.548	.9687217	1.061718
	exper	.9595555	.0292613	-1.35	0.176	.9038846	1.018655
pyperstudlep		.9960841	.0190104	-0.21	0.837	.9595126	1.034049
pyperstude~s		.9834924	.0140734	-1.16	0.245	.9562922	1.011466
pyperstudw~e		.9113888	.1510453	-0.56	0.576	.6586175	1.261171
pyperstudb~k		.9458279	.1641091	-0.32	0.748	.673165	1.328932
pyperstudl~o		.9272896	.1536699	-0.46	0.649	.6701234	1.283146
pyperstudm~e		1.012643	.0133608	0.95	0.341	.9867923	1.039172
	male	2.272857	.8692114	2.15	0.032	1.074103	4.80948
	age	1.004723	.0275605	0.17	0.864	.9521315	1.060219
	Asian	(omitted)					
	Black	9.326176	12.04616	1.73	0.084	.7417348	117.2623
	Latino	1.860464	1.055786	1.09	0.274	.6117543	5.658034
	Other	3289120	2.91e+09	0.02	0.986	0	.
	AsianMale	(omitted)					
	BlackMale	.1106237	186.8861	-0.00	0.999	0	.
	LatinoMale	.6801617	.4890364	-0.54	0.592	.1661874	2.783724
	OtherMale	6.54e-13	1.26e-09	-0.01	0.988	0	.
3							
	tenure	.9986111	.0160417	-0.09	0.931	.9676597	1.030553
	exper	1.122713	.0320195	4.06	0.000	1.061678	1.187257
pyperstudlep		1.000348	.0140481	0.02	0.980	.9731898	1.028264
pyperstude~s		1.043279	.0117259	3.77	0.000	1.020548	1.066516
pyperstudw~e		.924895	.1362588	-0.53	0.596	.6929317	1.23451
pyperstudb~k		.8957949	.1368302	-0.72	0.471	.6640333	1.208446
pyperstudl~o		.903743	.1329597	-0.69	0.491	.6773532	1.205798
pyperstudm~e		.9920624	.0124079	-0.64	0.524	.968039	1.016682
	male	1.660051	.4640872	1.81	0.070	.9597497	2.871343
	age	.9194849	.0248345	-3.11	0.002	.872076	.969471
	Asian	(omitted)					
	Black	4.263369	5.250163	1.18	0.239	.3815342	47.64008
	Latino	1.560769	.6590661	1.05	0.292	.6821844	3.570883
	Other	3347369	2.96e+09	0.02	0.986	0	.
	AsianMale	(omitted)					
	BlackMale	199560.6	1.70e+08	0.01	0.989	0	.
	LatinoMale	.4175616	.2493828	-1.46	0.144	.1295246	1.346135
	OtherMale	2.92e-13	4.20e-10	-0.02	0.984	0	.

-----> region = 19

Iteration 0: log likelihood = -804.41287  
Iteration 1: log likelihood = -738.39053  
Iteration 2: log likelihood = -736.04898  
Iteration 3: log likelihood = -735.92865  
Iteration 4: log likelihood = -735.90419  
Iteration 5: log likelihood = -735.89905  
Iteration 6: log likelihood = -735.8979  
Iteration 7: log likelihood = -735.89762  
Iteration 8: log likelihood = -735.89756  
Iteration 9: log likelihood = -735.89755

Multinomial logistic regression	Number of obs	=	749
	LR chi2(36)	=	137.03
	Prob > chi2	=	0.0000
Log likelihood = -735.89755	Pseudo R2	=	0.0852

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
--	-------	-----	-----------	---	------	----------------------	--

1		(base outcome)					
2							
	tenure	1.014821	.0137227	1.09	0.277	.9882783	1.042077
	exper	1.062744	.0194712	3.32	0.001	1.025259	1.101601
	pyperstudlep	1.013418	.0061102	2.21	0.027	1.001513	1.025465
	pyperstude~s	.9945156	.0087376	-0.63	0.531	.9775367	1.011789
	pyperstudw~e	1.024104	.0872424	0.28	0.780	.8666247	1.2102
	pyperstudb~k	1.038359	.0917314	0.43	0.670	.8732735	1.234653
	pyperstudl~o	1.016839	.0819526	0.21	0.836	.8682591	1.190845
	pyperstudm~e	.9959355	.0077495	-0.52	0.601	.980862	1.011241
	male	1.514912	.4213629	1.49	0.135	.8782769	2.613025
	age	.9731848	.0149596	-1.77	0.077	.9443018	1.002951
	Asian	2815105	4.26e+09	0.01	0.992	0	.
	Black	.8969244	.5947555	-0.16	0.870	.2445228	3.289972
	Latino	1.201751	.3012888	0.73	0.464	.7352102	1.964345
	Other	2805790	4.25e+09	0.01	0.992	0	.
	AsianMale	5.19e-13	9.43e-10	-0.02	0.988	0	.
	BlackMale	.8570149	.8150474	-0.16	0.871	.1328831	5.527224
	LatinoMale	.9708932	.3503816	-0.08	0.935	.4786134	1.969509
	OtherMale	7.85e-13	1.65e-09	-0.01	0.989	0	.
3							
	tenure	1.048043	.0159444	3.08	0.002	1.017254	1.079764
	exper	1.112124	.0257295	4.59	0.000	1.062822	1.163714
	pyperstudlep	1.025023	.0071035	3.57	0.000	1.011195	1.039041
	pyperstude~s	.9999658	.0106837	-0.00	0.997	.9792438	1.021126
	pyperstudw~e	1.021702	.0926263	0.24	0.813	.8553723	1.220374
	pyperstudb~k	1.04326	.0977975	0.45	0.651	.8681588	1.253678
	pyperstudl~o	1.016614	.0864594	0.19	0.846	.8605269	1.201013
	pyperstudm~e	1.005786	.0077048	0.75	0.451	.9907978	1.021001
	male	.9421859	.3118588	-0.18	0.857	.4924852	1.80252
	age	.9692745	.0192754	-1.57	0.117	.9322223	1.007799
	Asian	.7720164	1935.852	-0.00	1.000	0	.
	Black	.498934	.4359378	-0.80	0.426	.0900151	2.765481
	Latino	.9519486	.2679453	-0.17	0.861	.5483091	1.652728
	Other	1.119068	2806.094	0.00	1.000	0	.
	AsianMale	2.078626	5212.214	0.00	1.000	0	.
	BlackMale	3.227271	3.729917	1.01	0.311	.3350113	31.08933
	LatinoMale	1.13236	.4874093	0.29	0.773	.4870751	2.632527
	OtherMale	8.68e-06	.0265016	-0.00	0.997	0	.

-----> region = 20

Iteration 0: log likelihood = -2145.4814  
Iteration 1: log likelihood = -2023.1452  
Iteration 2: log likelihood = -2018.5162  
Iteration 3: log likelihood = -2018.2692  
Iteration 4: log likelihood = -2018.2204  
Iteration 5: log likelihood = -2018.2101  
Iteration 6: log likelihood = -2018.2078  
Iteration 7: log likelihood = -2018.2072  
Iteration 8: log likelihood = -2018.2071  
Iteration 9: log likelihood = -2018.2071

Multinomial logistic regression	Number of obs	=	2126
	LR chi2(36)	=	254.55
	Prob > chi2	=	0.0000
Log likelihood = -2018.2071	Pseudo R2	=	0.0593

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

-----						
2	tenure	1.003445	.008425	0.41	0.682	.9870676
	exper	.9962515	.0106268	-0.35	0.725	.9756394
	pyperstudlep	.9827673	.0074699	-2.29	0.022	.9682351
	pyperstude~s	1.000429	.0045876	0.09	0.925	.991478
	pyperstudw~e	.9304018	.0377562	-1.78	0.075	.8592673
	pyperstudb~k	.9274918	.0397623	-1.76	0.079	.8527435
	pyperstudl~o	.9314045	.0378117	-1.75	0.080	.8601667
	pyperstudm~e	1.00206	.0039031	0.53	0.597	.994439
	male	1.542135	.2211904	3.02	0.003	1.164216
	age	.967398	.0088941	-3.61	0.000	.950122
	Asian	8.25e-07	.0004957	-0.02	0.981	0
	Black	1.124564	.3490047	0.38	0.705	.612092
	Latino	1.247139	.2026248	1.36	0.174	.9070236
	Other	2.169207	1.797534	0.93	0.350	.4275085
	AsianMale	2074025	1.25e+09	0.02	0.981	0
	BlackMale	1.107249	.5328281	0.21	0.832	.431153
	LatinoMale	1.428741	.339189	1.50	0.133	.8971701
	OtherMale	.4062455	.6111956	-0.60	0.549	.0212891
-----						
3	tenure	1.024478	.0083507	2.97	0.003	1.008241
	exper	1.081761	.0143639	5.92	0.000	1.053972
	pyperstudlep	1.020186	.0067314	3.03	0.002	1.007078
	pyperstude~s	1.022299	.0052424	4.30	0.000	1.012075
	pyperstudw~e	1.048215	.049684	0.99	0.320	.9552226
	pyperstudb~k	1.016989	.0505491	0.34	0.735	.9225871
	pyperstudl~o	1.018309	.0482107	0.38	0.702	.9280693
	pyperstudm~e	1.011631	.0039386	2.97	0.003	1.003941
	male	1.402866	.2180824	2.18	0.029	1.03441
	age	.9538702	.0115911	-3.89	0.000	.9314205
	Asian	1.55e-06	.0009738	-0.02	0.983	0
	Black	1.171236	.3505904	0.53	0.597	.6514046
	Latino	.9706111	.1664639	-0.17	0.862	.6935225
	Other	1.677882	2.000804	0.43	0.664	.1620837
	AsianMale	1.28608	1476.465	0.00	1.000	0
	BlackMale	1.132549	.5770472	0.24	0.807	.417215
	LatinoMale	1.120282	.2940878	0.43	0.665	.6696941
	OtherMale	1.515423	2.296183	0.27	0.784	.0777652
-----						

by pyschooltype:

```
. bysort pyschooltype: mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age A
> sian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert2001a==1,
rrr
```

-----> pyschooltype =

```
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -507.82875
Iteration 1: log likelihood = -471.66774
Iteration 2: log likelihood = -470.43273
Iteration 3: log likelihood = -470.32943
Iteration 4: log likelihood = -470.30677
Iteration 5: log likelihood = -470.30208
Iteration 6: log likelihood = -470.3013
Iteration 7: log likelihood = -470.30112
Iteration 8: log likelihood = -470.30108
Iteration 9: log likelihood = -470.30107
```

```
Multinomial logistic regression      Number of obs   =      475
                                      LR chi2(34)       =      75.06
                                      Prob > chi2      =      0.0001
Log likelihood = -470.30107          Pseudo R2       =      0.0739
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.039103	.0185927	2.14	0.032	1.003294	1.076191
	exper	1.007311	.0236646	0.31	0.757	.9619805	1.054777
pyperstudlep		1.007424	.0114001	0.65	0.513	.9853263	1.030018
pyperstude~s		.9873429	.0107122	-1.17	0.240	.9665689	1.008563
pyperstudw~e		.9910214	.0451715	-0.20	0.843	.9063263	1.083631
pyperstudb~k		1.005053	.0500779	0.10	0.919	.9115427	1.108156
pyperstudl~o		1.000772	.0468655	0.02	0.987	.9130064	1.096974
pyperstudm~e		.9606394	.0172626	-2.23	0.025	.9273942	.9950764
	male	2.191694	.6371114	2.70	0.007	1.23977	3.874528
	age	.9509162	.0193347	-2.48	0.013	.9137661	.9895766
	Asian	3.52e-07	.0004699	-0.01	0.991	0	.
	Black	1.415717	.6854193	0.72	0.473	.5481096	3.656669
	Latino	1.48205	.6615691	0.88	0.378	.6178696	3.554912
	Other	6.07e-07	.0011599	-0.01	0.994	0	.
	AsianMale	.9774726	2277.439	-0.00	1.000	0	.
	BlackMale	.5208875	.395629	-0.86	0.390	.1175525	2.308107
	LatinoMale	1.04389	.6438535	0.07	0.944	.3116364	3.496723
	OtherMale	(omitted)					
3							
	tenure	1.016583	.0173632	0.96	0.336	.9831152	1.05119
	exper	1.041966	.0247791	1.73	0.084	.9945147	1.091682
pyperstudlep		1.022602	.0126531	1.81	0.071	.9981011	1.047705
pyperstude~s		1.006008	.010255	0.59	0.557	.9861079	1.026309
pyperstudw~e		.9391152	.0403886	-1.46	0.144	.8631994	1.021707
pyperstudb~k		.9257625	.0444724	-1.61	0.108	.8425758	1.017162
pyperstudl~o		.9142723	.0405727	-2.02	0.043	.8381114	.9973541
pyperstudm~e		.9974222	.010486	-0.25	0.806	.9770803	1.018188
	male	1.7768	.5231355	1.95	0.051	.997753	3.164129
	age	.9738976	.0211267	-1.22	0.223	.9333578	1.016198
	Asian	4.91e-07	.000706	-0.01	0.992	0	.
	Black	.6011805	.3762817	-0.81	0.416	.1762925	2.050104
	Latino	2.290212	1.059789	1.79	0.073	.9246653	5.672399
	Other	2.19e-07	.000445	-0.01	0.994	0	.
	AsianMale	1.064008	2647.485	0.00	1.000	0	.
	BlackMale	1.249796	1.144214	0.24	0.808	.2077528	7.518503
	LatinoMale	.739992	.5106196	-0.44	0.663	.1913679	2.861442
	OtherMale	(omitted)					

-----> pyschooltype = B

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -629.64743
Iteration 1: log likelihood = -596.40442
Iteration 2: log likelihood = -592.70771
Iteration 3: log likelihood = -592.50462
Iteration 4: log likelihood = -592.49465
Iteration 5: log likelihood = -592.49266
Iteration 6: log likelihood = -592.49225
Iteration 7: log likelihood = -592.49215
Iteration 8: log likelihood = -592.49212
```

Multinomial logistic regression

```
Number of obs    =      661
LR chi2(34)      =      74.31
Prob > chi2      =      0.0001
Pseudo R2       =      0.0590
```

Log likelihood = -592.49212

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.023633	.0203631	1.17	0.240	.9844898	1.064332
	exper	.9841556	.0252757	-0.62	0.534	.9358422	1.034963
pyperstudlep		1.01351	.0119917	1.13	0.257	.9902773	1.037288
pyperstude~s		.9865618	.0073514	-1.82	0.069	.9722581	1.001076
pyperstudw~e		.9761108	.0407489	-0.58	0.562	.8994245	1.059335
pyperstudb~k		.994914	.0433456	-0.12	0.907	.9134842	1.083603
pyperstudl~o		.9889478	.041654	-0.26	0.792	.9105863	1.074053
pyperstudm~e		.9989857	.0044889	-0.23	0.821	.9902262	1.007823
	male	1.214634	.3959353	0.60	0.551	.6411779	2.300977
	age	.9661207	.0212684	-1.57	0.117	.925322	1.008718
	Asian	.0000302	.0129705	-0.02	0.981	0	.
	Black	2.128149	1.096359	1.47	0.143	.7753337	5.841382
	Latino	1.798976	1.130096	0.93	0.350	.5251826	6.162263
	Other	5.40e+07	1.70e+11	0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	1.441227	.964411	0.55	0.585	.3882779	5.34961
	LatinoMale	1.469521	1.139246	0.50	0.620	.3215748	6.715362
	OtherMale	2.87e-07	.0009033	-0.00	0.996	0	.
3							
	tenure	.9851633	.0118533	-1.24	0.214	.962203	1.008672
	exper	1.033713	.0168893	2.03	0.042	1.001135	1.067351
pyperstudlep		1.011626	.0092063	1.27	0.204	.9937418	1.029832
pyperstude~s		.9967983	.0050275	-0.64	0.525	.9869931	1.006701
pyperstudw~e		1.049021	.039268	1.28	0.201	.9748121	1.128878
pyperstudb~k		1.049037	.040527	1.24	0.215	.9725384	1.131553
pyperstudl~o		1.044602	.0392014	1.16	0.245	.9705263	1.124332
pyperstudm~e		1.006183	.0032185	1.93	0.054	.999895	1.012511
	male	1.521073	.3063855	2.08	0.037	1.024931	2.257383
	age	.9864987	.0149131	-0.90	0.369	.9576984	1.016165
	Asian	1.807918	2.604225	0.41	0.681	.1074141	30.4296
	Black	.9944014	.459529	-0.01	0.990	.401983	2.45989
	Latino	.4355326	.2968218	-1.22	0.223	.1145287	1.656254
	Other	1.330428	6915.592	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.6715775	.4386445	-0.61	0.542	.1866935	2.415812
	LatinoMale	4.265544	3.300706	1.87	0.061	.9360717	19.43747
	OtherMale	1.257204	6534.974	0.00	1.000	0	.

-----> pyschooltype = E

Iteration 0: log likelihood = -9634.2184  
Iteration 1: log likelihood = -9023.5899  
Iteration 2: log likelihood = -9017.2058  
Iteration 3: log likelihood = -9017.2007  
Iteration 4: log likelihood = -9017.2007

Multinomial logistic regression	Number of obs	=	8982
	LR chi2(36)	=	1234.04
	Prob > chi2	=	0.0000
Log likelihood = -9017.2007	Pseudo R2	=	0.0640

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.002776	.0046131	0.60	0.547	.9937751	1.011858
	exper	1.000384	.0060941	0.06	0.950	.9885105	1.012399

pyperstudlep		1.012438	.0021688	5.77	0.000	1.008196	1.016698
pyperstude~s		.9812818	.0023214	-7.99	0.000	.9767425	.9858423
pyperstudw~e		.9899991	.0066845	-1.49	0.137	.976984	1.003188
pyperstudb~k		.9958713	.0073099	-0.56	0.573	.9816467	1.010302
pyperstudl~o		.9960464	.0069932	-0.56	0.573	.9824339	1.009847
pyperstudm~e		1.014898	.0038813	3.87	0.000	1.007319	1.022533
male		1.87265	.1721171	6.83	0.000	1.563946	2.242288
age		.9745343	.0047891	-5.25	0.000	.9651929	.983966
Asian		.9842148	.4272507	-0.04	0.971	.4203205	2.304619
Black		1.391118	.1506765	3.05	0.002	1.125039	1.720127
Latino		1.103067	.1020857	1.06	0.289	.9200803	1.322447
Other		.6964304	.2758524	-0.91	0.361	.3204213	1.51368
AsianMale		.9830958	.7478692	-0.02	0.982	.2213413	4.366458
BlackMale		.8827888	.1919108	-0.57	0.566	.5765179	1.351764
LatinoMale		.6404233	.1063571	-2.68	0.007	.4624933	.8868064
OtherMale		.4849176	.3776596	-0.93	0.353	.1053761	2.231484
-----							
3							
tenure		1.014572	.0037725	3.89	0.000	1.007204	1.021993
exper		1.067187	.0059078	11.75	0.000	1.055671	1.07883
pyperstudlep		1.001921	.0019386	0.99	0.321	.9981285	1.005728
pyperstude~s		1.003191	.0021116	1.51	0.130	.9990611	1.007339
pyperstudw~e		1.009679	.006873	1.42	0.157	.9962979	1.02324
pyperstudb~k		.997118	.0073419	-0.39	0.695	.9828315	1.011612
pyperstudl~o		1.000382	.0070771	0.05	0.957	.9866066	1.014349
pyperstudm~e		1.002841	.003634	0.78	0.434	.9957433	1.009989
male		3.020523	.2365582	14.11	0.000	2.590709	3.521646
age		.9833961	.0046392	-3.55	0.000	.9743454	.9925309
Asian		.4758559	.2765829	-1.28	0.201	.1523104	1.486694
Black		1.088376	.1141245	0.81	0.419	.8861844	1.3367
Latino		1.144568	.0972483	1.59	0.112	.9689899	1.351961
Other		.4435213	.1856172	-1.94	0.052	.19529	1.007277
AsianMale		.637397	.6127144	-0.47	0.639	.0968658	4.194205
BlackMale		.6069128	.1259066	-2.41	0.016	.4041485	.9114054
LatinoMale		.5115076	.0751909	-4.56	0.000	.3834659	.6823034
OtherMale		.3241209	.2601414	-1.40	0.160	.0672245	1.56274
-----							

-----> pyschooltype = M

Iteration 0: log likelihood = -5433.1562  
Iteration 1: log likelihood = -5123.8548  
Iteration 2: log likelihood = -5118.047  
Iteration 3: log likelihood = -5117.8195  
Iteration 4: log likelihood = -5117.7703  
Iteration 5: log likelihood = -5117.7598  
Iteration 6: log likelihood = -5117.7576  
Iteration 7: log likelihood = -5117.7572  
Iteration 8: log likelihood = -5117.757  
Iteration 9: log likelihood = -5117.757

Multinomial logistic regression	Number of obs	=	5121
	LR chi2(36)	=	630.80
	Prob > chi2	=	0.0000
Log likelihood = -5117.757	Pseudo R2	=	0.0581

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.016585	.0052446	3.19	0.001	1.006358 1.026916
exper		.9945893	.0066263	-0.81	0.415	.9816863 1.007662
pyperstudlep		1.00923	.0038548	2.41	0.016	1.001703 1.016814
pyperstude~s		.9936716	.00288	-2.19	0.028	.9880429 .9993323



pyperstudw~e		.9775266	.0088853	-2.50	0.012	.9602661	.9950975
pyperstudb~k		.9816488	.0097293	-1.87	0.062	.9627637	1.000904
pyperstudl~o		.9770518	.0093654	-2.42	0.015	.9588674	.9955811
pyperstudm~e		1.00878	.0036061	2.45	0.014	1.001737	1.015873
male		1.643337	.136917	5.96	0.000	1.39575	1.934843
age		.9638101	.0054145	-6.56	0.000	.9532561	.974481
Asian		.420573	.3404927	-1.07	0.285	.0860429	2.055738
Black		1.271657	.1684732	1.81	0.070	.980844	1.648693
Latino		1.856898	.2495344	4.61	0.000	1.426926	2.416432
Other		.6596722	.3006016	-0.91	0.361	.2700535	1.611412
AsianMale		1.060207	1.226011	0.05	0.960	.1099185	10.22611
BlackMale		1.167089	.2276903	0.79	0.428	.7962347	1.710673
LatinoMale		1.103119	.1936995	0.56	0.576	.781912	1.556278
OtherMale		1.333617	.8684906	0.44	0.658	.3721376	4.779237

3							
tenure		1.007394	.005105	1.45	0.146	.9974381	1.01745
exper		1.074575	.0085165	9.08	0.000	1.058012	1.091397
pyperstudlep		.999455	.0047056	-0.12	0.908	.9902746	1.00872
pyperstude~s		1.007754	.0032564	2.39	0.017	1.001392	1.014157
pyperstudw~e		1.027081	.012544	2.19	0.029	1.002787	1.051963
pyperstudb~k		1.008672	.0132943	0.66	0.512	.982949	1.035067
pyperstudl~o		1.011958	.0129136	0.93	0.352	.9869618	1.037587
pyperstudm~e		1.006784	.0042649	1.60	0.111	.9984591	1.015178
male		2.270421	.2045529	9.10	0.000	1.902907	2.708913
age		.954056	.0069531	-6.45	0.000	.9405251	.9677816
Asian		2.68e-06	.001158	-0.03	0.976	0	.
Black		1.155263	.1988262	0.84	0.402	.8244894	1.618739
Latino		1.577792	.2594034	2.77	0.006	1.143153	2.177685
Other		.213621	.2213704	-1.49	0.136	.0280262	1.628261
AsianMale		543267.9	2.35e+08	0.03	0.976	0	.
BlackMale		1.23726	.2898412	0.91	0.363	.7817326	1.958231
LatinoMale		.9096542	.1870497	-0.46	0.645	.6079195	1.361152
OtherMale		3.885109	4.451044	1.18	0.236	.4113477	36.6942

-----> pyschooltype = S

```
Iteration 0: log likelihood = -6615.7138
Iteration 1: log likelihood = -6279.478
Iteration 2: log likelihood = -6262.4275
Iteration 3: log likelihood = -6262.1018
Iteration 4: log likelihood = -6262.0331
Iteration 5: log likelihood = -6262.0162
Iteration 6: log likelihood = -6262.0128
Iteration 7: log likelihood = -6262.0121
Iteration 8: log likelihood = -6262.0119
Iteration 9: log likelihood = -6262.0118
Iteration 10: log likelihood = -6262.0118
```

Multinomial logistic regression	Number of obs	=	6749
	LR chi2(36)	=	707.40
	Prob > chi2	=	0.0000
Log likelihood = -6262.0118	Pseudo R2	=	0.0535

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.006823	.0039734	1.72	0.085	.9990652 1.014641
exper		.9974584	.005473	-0.46	0.643	.986789 1.008243
pyperstudlep		1.017066	.0041328	4.16	0.000	1.008998 1.025198
pyperstude~s		.9913631	.0023124	-3.72	0.000	.9868412 .9959056
pyperstudw~e		.9821203	.0069566	-2.55	0.011	.9685797 .9958501

pyperstudb~k		.9877386	.0075549	-1.61	0.107	.9730417	1.002658
pyperstudl~o		.9837837	.0073386	-2.19	0.028	.9695049	.9982727
pyperstudm~e		.99887	.0023531	-0.48	0.631	.9942687	1.003493
male		1.089677	.0763532	1.23	0.220	.9498489	1.25009
age		.970447	.0049296	-5.91	0.000	.9608331	.9801572
Asian		.6916668	.4671382	-0.55	0.585	.1840823	2.598853
Black		1.303896	.1684489	2.05	0.040	1.012225	1.679611
Latino		1.418033	.1793526	2.76	0.006	1.106691	1.816963
Other		.7963767	.4694272	-0.39	0.699	.2508272	2.528497
AsianMale		2.046738	1.774552	0.83	0.409	.3741605	11.19609
BlackMale		1.917391	.3301055	3.78	0.000	1.368247	2.686933
LatinoMale		1.670229	.2585077	3.31	0.001	1.233196	2.262141
OtherMale		2.187924	1.571718	1.09	0.276	.5352573	8.943383
-----							
3							
tenure		.9930059	.0045463	-1.53	0.125	.9841352	1.001957
exper		1.055733	.0078075	7.33	0.000	1.040541	1.071147
pyperstudlep		1.002751	.0057381	0.48	0.631	.9915677	1.014061
pyperstude~s		1.015057	.0027777	5.46	0.000	1.009628	1.020516
pyperstudw~e		1.043761	.0131965	3.39	0.001	1.018214	1.069948
pyperstudb~k		1.026941	.0137004	1.99	0.046	1.000437	1.054147
pyperstudl~o		1.025107	.0133552	1.90	0.057	.9992623	1.051619
pyperstudm~e		1.021752	.0022064	9.96	0.000	1.017437	1.026086
male		2.138103	.2006325	8.10	0.000	1.778913	2.56982
age		.9647532	.0069662	-4.97	0.000	.9511959	.9785037
Asian		.774942	.8238682	-0.24	0.810	.0964554	6.226036
Black		.8655475	.1941949	-0.64	0.520	.5575896	1.343591
Latino		1.313064	.2470658	1.45	0.148	.9080803	1.898663
Other		2.091599	1.267188	1.22	0.223	.6379414	6.857662
AsianMale		2.80e-07	.0003838	-0.01	0.991	0	.
BlackMale		1.586627	.4331897	1.69	0.091	.9291266	2.709409
LatinoMale		1.035948	.2262742	0.16	0.872	.6751749	1.589495
OtherMale		1.106001	.825309	0.14	0.893	.2562008	4.77453
-----							

## 2001-02

```
. mlogit admin tenure exper charter pyperstudlep pyperstudecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian Black La
> tino Other AsianMale BlackMale LatinoMale OtherMale if validcert2002a==1, rrr
```

```
Iteration 0: log likelihood = -23853.169
Iteration 1: log likelihood = -22777.084
Iteration 2: log likelihood = -22760.604
Iteration 3: log likelihood = -22760.563
Iteration 4: log likelihood = -22760.563
```

```
Multinomial logistic regression      Number of obs   =      22452
                                     LR chi2(38)       =      2185.21
                                     Prob > chi2       =      0.0000
Log likelihood = -22760.563          Pseudo R2       =      0.0458
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.006429	.0024759	2.60	0.009	1.001587 1.011293
exper		.9995665	.0032656	-0.13	0.894	.9931865 1.005988
charter		.2492319	.135299	-2.56	0.010	.0860037 .7222543
pyperstudlep		1.00921	.0014202	6.51	0.000	1.00643 1.011997
pyperstude~s		.9920884	.0012058	-6.54	0.000	.9897278 .9944546
pyperstudw~e		.9791736	.0039466	-5.22	0.000	.9714689 .9869394
pyperstudb~k		.9855781	.0043024	-3.33	0.001	.9771815 .9940468
pyperstudl~o		.9828585	.0041166	-4.13	0.000	.9748232 .99096
pyperstudm~e		.9994773	.0014311	-0.37	0.715	.9966764 1.002286

male		1.263669	.0526795	5.61	0.000	1.164525	1.371255
age		.9698906	.002717	-10.91	0.000	.9645801	.9752305
Asian		.6820608	.220952	-1.18	0.238	.361475	1.286968
Black		1.186183	.0795683	2.55	0.011	1.040049	1.35285
Latino		1.276931	.0789017	3.96	0.000	1.131284	1.441329
Other		.7757512	.1903223	-1.03	0.301	.4796111	1.254746
AsianMale		3.50857	1.761329	2.50	0.012	1.311658	9.385118
BlackMale		1.58399	.1633924	4.46	0.000	1.294044	1.938902
LatinoMale		1.250558	.1082416	2.58	0.010	1.055428	1.481765
OtherMale		1.518227	.5357127	1.18	0.237	.7603032	3.031704
-----							
3							
tenure		1.01341	.0022874	5.90	0.000	1.008936	1.017903
exper		1.053444	.003629	15.11	0.000	1.046355	1.060581
charter		2.635046	.7835185	3.26	0.001	1.471254	4.719424
pyperstudlep		1.012443	.001441	8.69	0.000	1.009622	1.015271
pyperstude~s		1.019614	.0012575	15.75	0.000	1.017152	1.022081
pyperstudw~e		1.009426	.0047659	1.99	0.047	1.000129	1.018811
pyperstudb~k		.9900103	.0050494	-1.97	0.049	.980163	.9999566
pyperstudl~o		.9862124	.0048351	-2.83	0.005	.9767812	.9957346
pyperstudm~e		1.002629	.0014327	1.84	0.066	.9998246	1.005441
male		1.364057	.0557732	7.59	0.000	1.259008	1.477787
age		.9748522	.0030371	-8.18	0.000	.9689177	.980823
Asian		.5836579	.2347449	-1.34	0.181	.2653436	1.283832
Black		.8429512	.0633563	-2.27	0.023	.7274883	.9767397
Latino		1.189777	.0767317	2.69	0.007	1.048503	1.350087
Other		.493113	.1462318	-2.38	0.017	.2757555	.8817972
AsianMale		1.711785	1.059044	0.87	0.385	.5091308	5.755316
BlackMale		1.551649	.1796477	3.79	0.000	1.236638	1.946904
LatinoMale		1.041761	.0948232	0.45	0.653	.871545	1.24522
OtherMale		2.209522	.8849898	1.98	0.048	1.007773	4.844332

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=22452)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		35.304	2	0.000
exper		267.052	2	0.000
charter		27.063	2	0.000
pyperstudlep		88.085	2	0.000
pyperstude~s		416.516	2	0.000
pyperstudw~e		44.401	2	0.000
pyperstudb~k		11.796	2	0.003
pyperstudl~o		19.230	2	0.000
pyperstudm~e		4.374	2	0.112
male		68.259	2	0.000
age		148.323	2	0.000
Asian		2.707	2	0.258
Black		17.138	2	0.000
Latino		17.640	2	0.000
Other		6.527	2	0.038
AsianMale		6.499	2	0.039
BlackMale		25.131	2	0.000
LatinoMale		7.039	2	0.030
OtherMale		4.409	2	0.110

\*\*\*\* Wald tests for independent variables (N=22452)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
--	--	------	----	--------

tenure		35.151	2	0.000
exper		252.363	2	0.000
charter		22.307	2	0.000
pyperstudlep		86.785	2	0.000
pyperstude~s		402.238	2	0.000
pyperstudw~e		44.853	2	0.000
pyperstudb~k		11.741	2	0.003
pyperstudl~o		19.075	2	0.000
pyperstudm~e		4.442	2	0.109
male		68.053	2	0.000
age		146.031	2	0.000
Asian		2.619	2	0.270
Black		17.081	2	0.000
Latino		17.636	2	0.000
Other		5.954	2	0.051
AsianMale		6.252	2	0.044
BlackMale		24.762	2	0.000
LatinoMale		7.042	2	0.030
OtherMale		4.322	2	0.115

\*\*\*\* Hausman tests of IIA assumption (N=22452)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-13.309	19	---	---
3		-23.101	19	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=22452)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-5144.238	-5127.540	33.397	20	0.031	against Ho
3		-5281.821	-5273.946	15.749	20	0.732	for Ho

\*\*\*\* Wald tests for combining alternatives (N=22452)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1435.768	19	0.000
2-	1	608.293	19	0.000
3-	1	1077.436	19	0.000

\*\*\*\* LR tests for combining alternatives (N=22452)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1608.034	19	0.000
2-	1	635.102	19	0.000
3-	1	1177.128	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper charter pyperstudlep pyperstudcodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age
> Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if
validcert2002a==1, rrr
```

```
-----> region = 1
```

```
Iteration 0: log likelihood = -1611.3439
Iteration 1: log likelihood = -1518.2811
Iteration 2: log likelihood = -1515.72
Iteration 3: log likelihood = -1515.4035
Iteration 4: log likelihood = -1515.3531
Iteration 5: log likelihood = -1515.3406
Iteration 6: log likelihood = -1515.3378
Iteration 7: log likelihood = -1515.3372
Iteration 8: log likelihood = -1515.3371
Iteration 9: log likelihood = -1515.3371
```

```
Multinomial logistic regression      Number of obs   =      1489
                                      LR chi2(38)        =      192.01
                                      Prob > chi2         =      0.0000
Log likelihood = -1515.3371          Pseudo R2        =      0.0596
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.01484	.0097334	1.54	0.125	.9959415	1.034098
	exper	1.030173	.015149	2.02	0.043	1.000906	1.060297
	charter	.7475102	2235.383	-0.00	1.000	0	.
	pyperstudlep	1.003964	.0034931	1.14	0.256	.9971407	1.010834
	pyperstude~s	1.001412	.0062425	0.23	0.821	.9892512	1.013722
	pyperstudw~e	.9801592	.1117841	-0.18	0.861	.7838261	1.22567
	pyperstudb~k	.798724	.1539529	-1.17	0.244	.5474303	1.165372
	pyperstudl~o	.9623729	.1020265	-0.36	0.718	.7818127	1.184634
	pyperstudm~e	.9980491	.0052226	-0.37	0.709	.9878652	1.008338
	male	1.896037	.5394606	2.25	0.025	1.085583	3.311546
	age	.9403391	.0121359	-4.77	0.000	.9168515	.9644284
	Asian	4.11e-07	.000622	-0.01	0.992	0	.
	Black	4822984	7.83e+09	0.01	0.992	0	.
	Latino	1.015835	.2213715	0.07	0.943	.6627184	1.557104
	Other	1.38e-06	.0008572	-0.02	0.983	0	.
	AsianMale	3.44e+12	7.63e+15	0.01	0.990	0	.
	BlackMale	2.03e-07	.0003302	-0.01	0.992	0	.
	LatinoMale	.9493755	.3016261	-0.16	0.870	.5093342	1.769592
	OtherMale	2.84e+11	2.89e+14	0.03	0.979	0	.
3							
	tenure	1.033015	.0100263	3.35	0.001	1.01355	1.052855
	exper	1.07801	.0181223	4.47	0.000	1.043069	1.11412
	charter	5226407	1.03e+10	0.01	0.994	0	.
	pyperstudlep	1.016169	.0036619	4.45	0.000	1.009017	1.023371
	pyperstude~s	1.002762	.0070263	0.39	0.694	.9890851	1.016629
	pyperstudw~e	.9606586	.1234455	-0.31	0.755	.7467735	1.235803
	pyperstudb~k	1.022375	.1317529	0.17	0.864	.7941748	1.316146
	pyperstudl~o	.9489595	.1126975	-0.44	0.659	.7518996	1.197665
	pyperstudm~e	1.003118	.0053843	0.58	0.562	.9926202	1.013727
	male	.6218887	.2319017	-1.27	0.203	.2994336	1.291591
	age	.9476418	.0143925	-3.54	0.000	.9198487	.9762746
	Asian	7.67e-07	.0012699	-0.01	0.993	0	.
	Black	2.670714	6922.544	0.00	1.000	0	.
	Latino	1.008183	.2316933	0.04	0.972	.6425726	1.581817
	Other	.6087267	.7309653	-0.41	0.679	.0578483	6.405512

AsianMale		4821579	1.48e+10	0.01	0.996	0	.
BlackMale		8.79e-07	.0024032	-0.01	0.996	0	.
LatinoMale		2.406733	.9680794	2.18	0.029	1.094063	5.29436
OtherMale		2495411	2.02e+09	0.02	0.985	0	.

-----> region = 2

```

Iteration 0: log likelihood = -812.55779
Iteration 1: log likelihood = -770.93265
Iteration 2: log likelihood = -769.7764
Iteration 3: log likelihood = -769.60916
Iteration 4: log likelihood = -769.58187
Iteration 5: log likelihood = -769.57907
Iteration 6: log likelihood = -769.5785
Iteration 7: log likelihood = -769.57837
Iteration 8: log likelihood = -769.57835
Iteration 9: log likelihood = -769.57834

```

Multinomial logistic regression	Number of obs	=	825
	LR chi2(38)	=	85.96
	Prob > chi2	=	0.0000
Log likelihood = -769.57834	Pseudo R2	=	0.0529

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.010298	.015164	0.68	0.495	.9810103 1.04046
exper		.9802698	.018748	-1.04	0.297	.9442045 1.017713
charter		3.949196	5.54946	0.98	0.328	.2514142 62.0337
pyperstudlep		.9754634	.0181131	-1.34	0.181	.9406005 1.011618
pyperstude~s		1.010523	.0066363	1.59	0.111	.9975999 1.023615
pyperstudw~e		.8829396	.068334	-1.61	0.108	.7586706 1.027564
pyperstudb~k		.8902011	.0689054	-1.50	0.133	.7648944 1.036036
pyperstudl~o		.8863913	.0645254	-1.66	0.098	.7685317 1.022326
pyperstudm~e		.9917384	.0079517	-1.03	0.301	.9762753 1.007447
male		1.473764	.4169364	1.37	0.170	.8464827 2.565889
age		.9794557	.015255	-1.33	0.183	.9500081 1.009816
Asian		7.45e+07	3.60e+11	0.00	0.997	0 .
Black		4.142313	3.166887	1.86	0.063	.925723 18.53552
Latino		1.169643	.3174797	0.58	0.564	.6870829 1.99112
Other		5.32e-07	.0014844	-0.01	0.996	0 .
AsianMale		5.92e-15	3.31e-11	-0.01	0.995	0 .
BlackMale		1.222222	1.307376	0.19	0.851	.1501912 9.946174
LatinoMale		1.037457	.4075486	0.09	0.925	.4803844 2.240532
OtherMale		2511678	7.01e+09	0.01	0.996	0 .
3						
tenure		1.000849	.0120825	0.07	0.944	.9774457 1.024813
exper		1.050736	.0185405	2.80	0.005	1.015019 1.08771
charter		3.328493	3.896991	1.03	0.304	.3354775 33.02418
pyperstudlep		.9914692	.0156459	-0.54	0.587	.9612731 1.022614
pyperstude~s		1.021397	.0064538	3.35	0.001	1.008826 1.034125
pyperstudw~e		1.20141	.0977953	2.25	0.024	1.024244 1.409222
pyperstudb~k		1.1253	.096652	1.37	0.169	.9509525 1.331613
pyperstudl~o		1.173704	.0910928	2.06	0.039	1.008082 1.366538
pyperstudm~e		.9972462	.007571	-0.36	0.716	.9825172 1.012196
male		1.134039	.2934932	0.49	0.627	.6828643 1.883311
age		.9638167	.0155899	-2.28	0.023	.9337404 .9948618
Asian		1.072877	9385.916	0.00	1.000	0 .
Black		3.03468	2.301651	1.46	0.143	.6863093 13.41856
Latino		.8685479	.2140495	-0.57	0.567	.5358204 1.407889
Other		7.92e-07	.0019403	-0.01	0.995	0 .

AsianMale		8.12e-07	.007374	-0.00	0.999	0	.
BlackMale		1.849271	1.973706	0.58	0.565	.2283087	14.97885
LatinoMale		1.720627	.6192068	1.51	0.132	.8498896	3.483461
OtherMale		1.835226	4885.659	0.00	1.000	0	.

-----> region = 3

note: charter omitted because of collinearity  
note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -317.82146  
Iteration 1: log likelihood = -292.04918  
Iteration 2: log likelihood = -291.32113  
Iteration 3: log likelihood = -291.22943  
Iteration 4: log likelihood = -291.21446  
Iteration 5: log likelihood = -291.211  
Iteration 6: log likelihood = -291.21019  
Iteration 7: log likelihood = -291.21002  
Iteration 8: log likelihood = -291.20999  
Iteration 9: log likelihood = -291.20999

Multinomial logistic regression	Number of obs	=	301
	LR chi2(30)	=	53.22
	Prob > chi2	=	0.0056
Log likelihood = -291.20999	Pseudo R2	=	0.0837

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.007911	.023365	0.34	0.734	.9631411 1.054762
exper		.9486909	.0278313	-1.80	0.073	.8956812 1.004838
charter		(omitted)				
pypersistudlep		1.02075	.042852	0.49	0.625	.9401239 1.10829
pypersistude~s		.9753335	.0149931	-1.62	0.104	.9463859 1.005167
pypersistudw~e		.9875219	.0461261	-0.27	0.788	.9011312 1.082195
pypersistudb~k		1.001177	.0475515	0.02	0.980	.9121841 1.098852
pypersistudl~o		1.016224	.0493372	0.33	0.740	.9239835 1.117674
pypersistudm~e		.9924341	.0201879	-0.37	0.709	.953645 1.032801
male		1.689253	.6352127	1.39	0.163	.8083811 3.52999
age		1.032089	.0305827	1.07	0.286	.9738557 1.093805
Asian		(omitted)				
Black		7.871441	7.46647	2.18	0.030	1.226438 50.51997
Latino		5.152935	3.570914	2.37	0.018	1.324901 20.04129
Other		.5551949	2086.926	-0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		4.58e-08	.0000963	-0.01	0.994	0 .
LatinoMale		.0669688	.0731488	-2.48	0.013	.0078726 .569672
OtherMale		(omitted)				
3						
tenure		.9868449	.018087	-0.72	0.470	.9520241 1.022939
exper		1.012942	.0270884	0.48	0.631	.9612168 1.06745
charter		(omitted)				
pypersistudlep		1.05569	.0376954	1.52	0.129	.9843345 1.132219
pypersistude~s		1.018835	.0127326	1.49	0.135	.9941827 1.044099
pypersistudw~e		1.095072	.0515616	1.93	0.054	.998536 1.200941
pypersistudb~k		1.069359	.0508952	1.41	0.159	.9741174 1.173912
pypersistudl~o		1.069109	.051605	1.38	0.166	.9726024 1.175192
pypersistudm~e		1.020543	.0158193	1.31	0.190	.9900036 1.052024
male		2.24367	.7197469	2.52	0.012	1.196471 4.20742
age		1.033117	.0281418	1.20	0.232	.9794064 1.089773

Asian		(omitted)					
Black		3.071093	2.874349	1.20	0.231	.4904809	19.22932
Latino		3.349012	2.327053	1.74	0.082	.8579488	13.07291
Other		8484786	1.81e+10	0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		.3130992	.5408453	-0.67	0.501	.0106004	9.247862
LatinoMale		.2767808	.2519673	-1.41	0.158	.0464777	1.648265
OtherMale		(omitted)					

-----

-----> region = 4

```

Iteration 0: log likelihood = -4824.1353
Iteration 1: log likelihood = -4553.2917
Iteration 2: log likelihood = -4544.4925
Iteration 3: log likelihood = -4544.3143
Iteration 4: log likelihood = -4544.2737
Iteration 5: log likelihood = -4544.2647
Iteration 6: log likelihood = -4544.2627
Iteration 7: log likelihood = -4544.2624
Iteration 8: log likelihood = -4544.2623
Iteration 9: log likelihood = -4544.2623

```

Multinomial logistic regression	Number of obs	=	4649
	LR chi2(38)	=	559.75
	Prob > chi2	=	0.0000
Log likelihood = -4544.2623	Pseudo R2	=	0.0580

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.015153	.0053457	2.86	0.004	1.00473 1.025685
exper		1.026374	.0074463	3.59	0.000	1.011882 1.041072
charter		.2805462	.3027971	-1.18	0.239	.0338292 2.326575
pyperstudlep		1.003864	.0036677	1.06	0.291	.9967012 1.011079
pyperstude~s		.991596	.0030775	-2.72	0.007	.9855824 .9976462
pyperstudw~e		.9939282	.0063394	-0.95	0.340	.9815806 1.006431
pyperstudb~k		1.001588	.0073945	0.21	0.830	.9871996 1.016187
pyperstudl~o		1.001152	.007195	0.16	0.873	.9871485 1.015353
pyperstudm~e		1.002342	.0035024	0.67	0.503	.9955007 1.00923
male		1.11265	.1016532	1.17	0.243	.9302326 1.330839
age		.9472423	.0058647	-8.75	0.000	.9358173 .9588069
Asian		.5936164	.3236369	-0.96	0.339	.2039082 1.728132
Black		.8170995	.0914521	-1.80	0.071	.6561544 1.017522
Latino		1.108602	.1666785	0.69	0.493	.825653 1.488518
Other		.3472256	.195579	-1.88	0.060	.1151231 1.047276
AsianMale		2.971438	2.691268	1.20	0.229	.5035273 17.53518
BlackMale		1.749513	.3023585	3.24	0.001	1.246835 2.454854
LatinoMale		1.541168	.3829149	1.74	0.082	.9470273 2.508057
OtherMale		1.656597	1.662229	0.50	0.615	.2318044 11.83892
-----						
3						
tenure		1.034713	.0060253	5.86	0.000	1.02297 1.04659
exper		1.06797	.0094806	7.41	0.000	1.049549 1.086714
charter		6.037089	3.743752	2.90	0.004	1.790507 20.35538
pyperstudlep		1.021877	.004317	5.12	0.000	1.013451 1.030374
pyperstude~s		1.022617	.003928	5.82	0.000	1.014947 1.030345
pyperstudw~e		1.003906	.0080684	0.49	0.628	.9882166 1.019845
pyperstudb~k		.9879707	.0094201	-1.27	0.204	.969679 1.006607
pyperstudl~o		.9742429	.0088914	-2.86	0.004	.9569711 .9918265
pyperstudm~e		1.002795	.0041379	0.68	0.499	.9947173 1.010938
male		1.081564	.1157077	0.73	0.464	.8769786 1.333875
age		.9631065	.0075134	-4.82	0.000	.9484926 .9779456



Asian		.3235551	.2609113	-1.40	0.162	.0666114	1.571621
Black		.4856879	.0688059	-5.10	0.000	.3679343	.6411274
Latino		1.122081	.1913503	0.68	0.499	.8032806	1.567404
Other		.3678253	.2475455	-1.49	0.137	.0983524	1.375619
AsianMale		1.97e-06	.0015486	-0.02	0.987	0	.
BlackMale		1.848109	.404869	2.80	0.005	1.202966	2.83924
LatinoMale		1.11618	.3431459	0.36	0.721	.6110128	2.039004
OtherMale		3.068326	3.312673	1.04	0.299	.369755	25.46179

-----> region = 5

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -578.25385  
Iteration 1: log likelihood = -521.20706  
Iteration 2: log likelihood = -519.19088  
Iteration 3: log likelihood = -519.06932  
Iteration 4: log likelihood = -519.03988  
Iteration 5: log likelihood = -519.03427  
Iteration 6: log likelihood = -519.03336  
Iteration 7: log likelihood = -519.03313  
Iteration 8: log likelihood = -519.03309  
Iteration 9: log likelihood = -519.03308

Multinomial logistic regression	Number of obs	=	574
	LR chi2(36)	=	118.44
	Prob > chi2	=	0.0000
Log likelihood = -519.03308	Pseudo R2	=	0.1024

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.017395	.0175338	1.00	0.317	.9836031 1.052347
exper		.9508025	.0221245	-2.17	0.030	.9084132 .9951698
charter		1.772471	6219.558	0.00	1.000	0 .
pypersistudlep		.9963293	.0485021	-0.08	0.940	.9056612 1.096074
pypersistude~s		1.008172	.0090308	0.91	0.364	.9906265 1.026028
pypersistudw~e		.9767537	.0485476	-0.47	0.636	.8860899 1.076694
pypersistudb~k		.9671122	.0490318	-0.66	0.510	.8756321 1.06815
pypersistudl~o		.9553079	.0606415	-0.72	0.471	.8435491 1.081873
pypersistudm~e		.9959715	.014375	-0.28	0.780	.9681918 1.024548
male		2.591757	.7271453	3.39	0.001	1.49548 4.491672
age		1.010233	.0188459	0.55	0.585	.9739631 1.047854
Asian		7.64e-07	.0015275	-0.01	0.994	0 .
Black		2.471838	.9366588	2.39	0.017	1.176176 5.194785
Latino		7360201	7.81e+09	0.01	0.988	0 .
Other		9.64e-07	.0019273	-0.01	0.994	0 .
AsianMale		(omitted)				
BlackMale		.9140075	.456204	-0.18	0.857	.3436321 2.431116
LatinoMale		3.24e-07	.0003437	-0.01	0.989	0 .
OtherMale		1.74e+13	7.21e+16	0.01	0.994	0 .
-----						
3						
tenure		1.00301	.0143029	0.21	0.833	.9753647 1.031438
exper		1.064794	.0269336	2.48	0.013	1.013293 1.118914
charter		1.14e+07	2.12e+10	0.01	0.993	0 .
pypersistudlep		1.030045	.0422461	0.72	0.470	.950485 1.116265
pypersistude~s		1.03795	.0094007	4.11	0.000	1.019687 1.056539
pypersistudw~e		.9835536	.0415779	-0.39	0.695	.905347 1.068516
pypersistudb~k		.9518309	.0412666	-1.14	0.255	.8742909 1.036248
pypersistudl~o		.9376036	.0505961	-1.19	0.233	.8435012 1.042204
pypersistudm~e		1.012257	.0130783	0.94	0.346	.9869463 1.038218
male		4.082864	1.094664	5.25	0.000	2.414058 6.905292

age		.9758876	.0222511	-1.07	0.284	.9332364	1.020488
Asian		5.91e-07	.0010946	-0.01	0.994	0	.
Black		1.757374	.726581	1.36	0.173	.7815155	3.951764
Latino		6093563	6.47e+09	0.01	0.988	0	.
Other		3.49e-06	.0064645	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		.8409207	.4565059	-0.32	0.750	.2901804	2.436924
LatinoMale		2.08e-07	.0002203	-0.01	0.988	0	.
OtherMale		219309	1.48e+09	0.00	0.999	0	.

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -759.36333  
Iteration 1: log likelihood = -705.24512  
Iteration 2: log likelihood = -703.36105  
Iteration 3: log likelihood = -703.17871  
Iteration 4: log likelihood = -703.13813  
Iteration 5: log likelihood = -703.12928  
Iteration 6: log likelihood = -703.12748  
Iteration 7: log likelihood = -703.12707  
Iteration 8: log likelihood = -703.12697  
Iteration 9: log likelihood = -703.12695

Multinomial logistic regression	Number of obs	=	698
	LR chi2(36)	=	112.47
	Prob > chi2	=	0.0000
Log likelihood = -703.12695	Pseudo R2	=	0.0741

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9606889	.0133961	-2.88	0.004	.9347885 .9873068
exper		.9823901	.017755	-0.98	0.326	.9482 1.017813
charter		1.315692	2514.555	0.00	1.000	0 .
pyperstudlep		1.018687	.017736	1.06	0.288	.9845115 1.054049
pyperstude~s		1.003557	.0080084	0.45	0.656	.9879834 1.019377
pyperstudw~e		.9647706	.0419732	-0.82	0.410	.8859145 1.050646
pyperstudb~k		.9611143	.043369	-0.88	0.379	.879763 1.049988
pyperstudl~o		.9645266	.0421804	-0.83	0.409	.8852986 1.050845
pyperstudm~e		.9682024	.0142592	-2.19	0.028	.9406543 .9965572
male		1.172888	.2398994	0.78	0.436	.7855142 1.751294
age		.9845521	.0153803	-1.00	0.319	.954864 1.015163
Asian		4.36e-07	.0004877	-0.01	0.990	0 .
Black		1.13592	.6273952	0.23	0.818	.3847773 3.353407
Latino		.448388	.3451179	-1.04	0.297	.0991965 2.026804
Other		2.525999	3.643583	0.64	0.521	.149497 42.68093
AsianMale		(omitted)				
BlackMale		4.219382	3.817963	1.59	0.112	.7161895 24.8582
LatinoMale		13.70384	18.66936	1.92	0.055	.9488806 197.9123
OtherMale		221733.8	1.72e+08	0.02	0.987	0 .
3						
tenure		.9745614	.0123981	-2.03	0.043	.9505621 .9991666
exper		1.057332	.0219686	2.68	0.007	1.015139 1.101278
charter		5497856	8.96e+09	0.01	0.992	0 .
pyperstudlep		1.021883	.0180778	1.22	0.221	.9870586 1.057937
pyperstude~s		1.024434	.0078674	3.14	0.002	1.00913 1.039971
pyperstudw~e		.9535158	.0405858	-1.12	0.263	.8771968 1.036475
pyperstudb~k		.938794	.0410629	-1.44	0.149	.8616656 1.022826
pyperstudl~o		.9337156	.0397427	-1.61	0.107	.8589819 1.014951
pyperstudm~e		1.007081	.0077902	0.91	0.362	.9919276 1.022466

male		1.82674	.3838644	2.87	0.004	1.210064	2.757688
age		.9747921	.0182658	-1.36	0.173	.9396411	1.011258
Asian		9.10e-07	.0011963	-0.01	0.992	0	.
Black		1.215745	.6928082	0.34	0.732	.3979	3.71459
Latino		1.080854	.8471505	0.10	0.921	.2326022	5.022509
Other		1.536671	2.323904	0.28	0.776	.0793068	29.77497
AsianMale		(omitted)					
BlackMale		1.869948	1.725591	0.68	0.498	.3064357	11.41089
LatinoMale		2.519996	3.690188	0.63	0.528	.142867	44.44959
OtherMale		753127.5	5.86e+08	0.02	0.986	0	.

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1023.0313  
Iteration 1: log likelihood = -965.2145  
Iteration 2: log likelihood = -962.55333  
Iteration 3: log likelihood = -962.29178  
Iteration 4: log likelihood = -962.24501  
Iteration 5: log likelihood = -962.23473  
Iteration 6: log likelihood = -962.23227  
Iteration 7: log likelihood = -962.23174  
Iteration 8: log likelihood = -962.23163  
Iteration 9: log likelihood = -962.2316

Multinomial logistic regression	Number of obs	=	1005
	LR chi2(36)	=	121.60
	Prob > chi2	=	0.0000
Log likelihood = -962.2316	Pseudo R2	=	0.0594

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9890184	.0129025	-0.85	0.397	.9640505 1.014633
exper		.9985414	.01739	-0.08	0.933	.9650328 1.033214
charter		1.29e-06	.0007482	-0.02	0.981	0 .
pyperstudlep		.9508461	.0233464	-2.05	0.040	.9061716 .997723
pyperstude~s		.9886475	.007773	-1.45	0.146	.9735295 1.004
pyperstudw~e		.8804174	.0948485	-1.18	0.237	.7128326 1.087401
pyperstudb~k		.88602	.0961465	-1.12	0.265	.7162676 1.096003
pyperstudl~o		.9324676	.1028878	-0.63	0.526	.751126 1.15759
pyperstudm~e		.9930571	.008059	-0.86	0.391	.9773867 1.008979
male		1.289688	.2693694	1.22	0.223	.856445 1.942093
age		.9698492	.0152684	-1.94	0.052	.9403806 1.000241
Asian		2.01e-06	.0021461	-0.01	0.990	0 .
Black		1.434248	.4886877	1.06	0.290	.7355233 2.796739
Latino		.3978014	.4348212	-0.84	0.399	.0466929 3.389078
Other		1.43e-06	.0011676	-0.02	0.987	0 .
AsianMale		(omitted)				
BlackMale		2.093685	1.03114	1.50	0.134	.7974336 5.497032
LatinoMale		2.157978	3.041685	0.55	0.585	.13623 34.18388
OtherMale		2.47327	2805.666	0.00	0.999	0 .
3						
tenure		.99359	.0097214	-0.66	0.511	.9747179 1.012828
exper		1.020372	.0141035	1.46	0.145	.9931003 1.048392
charter		.2758271	.3063609	-1.16	0.246	.0312758 2.432571
pyperstudlep		1.024197	.0212121	1.15	0.248	.983455 1.066628
pyperstude~s		1.028424	.0060051	4.80	0.000	1.016721 1.040261
pyperstudw~e		.9200101	.0898075	-0.85	0.393	.7598045 1.113995
pyperstudb~k		.9055604	.0888589	-1.01	0.312	.7471238 1.097595
pyperstudl~o		.8928764	.0890145	-1.14	0.256	.7343981 1.085553

pyperstudm~e		1.000826	.0066182	0.12	0.901	.987938	1.013882
male		1.426383	.2357517	2.15	0.032	1.031692	1.972069
age		.9958835	.0127955	-0.32	0.748	.9711179	1.021281
Asian		1.646031	2.413933	0.34	0.734	.0929257	29.15682
Black		.9645636	.2991886	-0.12	0.907	.5251761	1.771564
Latino		.2010468	.2169993	-1.49	0.137	.0242413	1.667398
Other		1.09784	1.368503	0.07	0.940	.0953853	12.63562
AsianMale		(omitted)					
BlackMale		1.716897	.8113531	1.14	0.253	.6799741	4.33507
LatinoMale		9.29263	12.08966	1.71	0.087	.7256547	119.0001
OtherMale		4.570436	7.889864	0.88	0.379	.1550755	134.7014

-----> region = 8

note: charter omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -330.94087  
Iteration 1: log likelihood = -290.11561  
Iteration 2: log likelihood = -287.47242  
Iteration 3: log likelihood = -287.13051  
Iteration 4: log likelihood = -287.08417  
Iteration 5: log likelihood = -287.07426  
Iteration 6: log likelihood = -287.07208  
Iteration 7: log likelihood = -287.07173  
Iteration 8: log likelihood = -287.07164  
Iteration 9: log likelihood = -287.07163

Multinomial logistic regression	Number of obs	=	335
	LR chi2(32)	=	87.74
	Prob > chi2	=	0.0000
Log likelihood = -287.07163	Pseudo R2	=	0.1326

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.013509	.0283677	0.48	0.632	.9594066 1.070662
exper		.9970408	.0387016	-0.08	0.939	.9240007 1.075855
charter		(omitted)				
pyperstudlep		.9595475	.0989092	-0.40	0.689	.784017 1.174377
pyperstude~s		.9907468	.0178998	-0.51	0.607	.9562777 1.026458
pyperstudw~e		.6075271	.178846	-1.69	0.090	.3411818 1.081796
pyperstudb~k		.637772	.1852391	-1.55	0.121	.3609411 1.126924
pyperstudl~o		.6044091	.1762739	-1.73	0.084	.3412565 1.070486
pyperstudm~e		.9514355	.0373878	-1.27	0.205	.8809077 1.02761
male		2.799801	1.248509	2.31	0.021	1.168296 6.709675
age		.9616631	.0344948	-1.09	0.276	.8963764 1.031705
Asian		6.65e+07	4.85e+11	0.00	0.998	0 .
Black		3.233288	2.540435	1.49	0.135	.693178 15.08148
Latino		18.70953	31.37899	1.75	0.081	.6989671 500.8057
Other		1.02e-06	.0010958	-0.01	0.990	0 .
AsianMale		(omitted)				
BlackMale		.7768115	.8529861	-0.23	0.818	.0902911 6.683232
LatinoMale		(omitted)				
OtherMale		3.12e+12	5.00e+15	0.02	0.986	0 .
3						
tenure		1.014271	.0180282	0.80	0.425	.9795448 1.050228
exper		.9947415	.0246011	-0.21	0.831	.9476742 1.044147
charter		(omitted)				
pyperstudlep		.9971352	.0392197	-0.07	0.942	.9231541 1.077045
pyperstude~s		1.048405	.0118738	4.17	0.000	1.025389 1.071938

pyperstudw~e		.9149961	.1676747	-0.48	0.628	.6389031	1.310399
pyperstudb~k		.8951577	.1628275	-0.61	0.543	.6267113	1.278591
pyperstudl~o		.8796607	.1616079	-0.70	0.485	.6136712	1.260941
pyperstudm~e		.9974867	.0146848	-0.17	0.864	.9691164	1.026688
male		2.49946	.7266233	3.15	0.002	1.413811	4.418764
age		1.027925	.0245203	1.15	0.248	.9809726	1.077125
Asian		.5424714	5976.249	-0.00	1.000	0	.
Black		2.316375	1.390966	1.40	0.162	.7139512	7.515348
Latino		5.238779	6.566655	1.32	0.186	.4490288	61.12037
Other		1.82e-06	.0012244	-0.02	0.984	0	.
AsianMale		(omitted)					
BlackMale		.6825717	.6330801	-0.41	0.681	.1108321	4.203695
LatinoMale		(omitted)					
OtherMale		4.94e+11	6.75e+14	0.02	0.984	0	.

-----> region = 9

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -227.55953  
Iteration 1: log likelihood = -205.97113  
Iteration 2: log likelihood = -203.9793  
Iteration 3: log likelihood = -203.59109  
Iteration 4: log likelihood = -203.50623  
Iteration 5: log likelihood = -203.487  
Iteration 6: log likelihood = -203.48325  
Iteration 7: log likelihood = -203.4824  
Iteration 8: log likelihood = -203.48219  
Iteration 9: log likelihood = -203.48215  
Iteration 10: log likelihood = -203.48214

Multinomial logistic regression	Number of obs	=	215
	LR chi2(32)	=	48.15
	Prob > chi2	=	0.0333
Log likelihood = -203.48214	Pseudo R2	=	0.1058

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1						
tenure		1.016852	.0223168	0.76	0.446	.9740392 1.061546
exper		.9930252	.0368375	-0.19	0.850	.9233872 1.067915
charter		7679384	2.04e+10	0.01	0.995	0 .
pyperstudlep		.9203088	.0604052	-1.27	0.206	.8092157 1.046653
pyperstude~s		.9832794	.0138024	-1.20	0.230	.956596 1.010707
pyperstudw~e		.9738229	.1173112	-0.22	0.826	.7690248 1.233161
pyperstudb~k		.9880317	.1315485	-0.09	0.928	.7610974 1.28263
pyperstudl~o		.9885377	.1208463	-0.09	0.925	.7779218 1.256176
pyperstudm~e		1.013977	.0174957	0.80	0.421	.9802591 1.048854
male		.7353045	.2538526	-0.89	0.373	.3737687 1.446544
age		1.001777	.0312735	0.06	0.955	.9423195 1.064986
Asian		(omitted)				
Black		.3673396	.4812554	-0.76	0.445	.0281773 4.788898
Latino		3830823	5.79e+09	0.01	0.992	0 .
Other		4.61e-07	.0007607	-0.01	0.993	0 .
AsianMale		(omitted)				
BlackMale		6872893	7.50e+09	0.01	0.988	0 .
LatinoMale		1.02e-13	2.37e-10	-0.01	0.990	0 .
OtherMale		(omitted)				
2						
tenure		1.014853	.0318364	0.47	0.638	.9543348 1.07921
exper		.9034995	.0412501	-2.22	0.026	.8261626 .988076
charter		5.025313	23326.24	0.00	1.000	0 .

pyperstudlep		1.048137	.0730115	0.67	0.500	.9143758	1.201466
pyperstude~s		.9856008	.0172996	-0.83	0.409	.9522708	1.020097
pyperstudw~e		.7791515	.1071351	-1.81	0.070	.5950861	1.02015
pyperstudb~k		.7918172	.1183134	-1.56	0.118	.590797	1.061235
pyperstudl~o		.7844484	.109727	-1.74	0.083	.5963484	1.031879
pyperstudm~e		1.025597	.0214768	1.21	0.227	.9843559	1.068567
male		1.316371	.588776	0.61	0.539	.5478478	3.162984
age		1.047664	.0360925	1.35	0.177	.9792591	1.120846
Asian		(omitted)					
Black		3.390236	3.597576	1.15	0.250	.4236142	27.13247
Latino		1.30e+07	1.96e+10	0.01	0.991	0	.
Other		2.308224	3.40448	0.57	0.571	.1281773	41.56662
AsianMale		(omitted)					
BlackMale		917089.1	1.00e+09	0.01	0.990	0	.
LatinoMale		1.84e-14	5.17e-11	-0.01	0.991	0	.
OtherMale		(omitted)					

3 | (base outcome)

-----> region = 10

Iteration 0: log likelihood = -3386.5994  
 Iteration 1: log likelihood = -3194.3412  
 Iteration 2: log likelihood = -3191.1267  
 Iteration 3: log likelihood = -3191.1169  
 Iteration 4: log likelihood = -3191.1169

Multinomial logistic regression	Number of obs	=	3154
	LR chi2(38)	=	390.97
	Prob > chi2	=	0.0000
Log likelihood = -3191.1169	Pseudo R2	=	0.0577

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.009316	.0066155	1.41	0.157	.9964327 1.022366
exper		1.00203	.0087145	0.23	0.816	.9850944 1.019256
charter		.2576752	.2888126	-1.21	0.226	.0286426 2.318106
pyperstudlep		.999034	.0052096	-0.19	0.853	.9888754 1.009297
pyperstude~s		.994755	.0035246	-1.48	0.138	.9878707 1.001687
pyperstudw~e		.9836996	.0088591	-1.82	0.068	.9664884 1.001217
pyperstudb~k		.9840772	.0092935	-1.70	0.089	.9660297 1.002462
pyperstudl~o		.9895174	.0091232	-1.14	0.253	.9717969 1.007561
pyperstudm~e		1.002023	.0042962	0.47	0.637	.9936383 1.010479
male		1.353777	.1476739	2.78	0.005	1.09319 1.676481
age		.9601299	.0070644	-5.53	0.000	.9463833 .9740762
Asian		2.19095	1.713147	1.00	0.316	.4732179 10.14387
Black		1.463497	.2257397	2.47	0.014	1.081674 1.980099
Latino		1.679941	.3365217	2.59	0.010	1.134445 2.487738
Other		.8940275	.4838378	-0.21	0.836	.309521 2.58233
AsianMale		.8672513	.8792982	-0.14	0.888	.1188829 6.326601
BlackMale		1.481303	.3427648	1.70	0.089	.9411975 2.331347
LatinoMale		1.154803	.3798548	0.44	0.662	.6060634 2.20038
OtherMale		1.191509	.8390203	0.25	0.803	.2997162 4.736795
3						
tenure		1.022672	.0063168	3.63	0.000	1.010366 1.035128
exper		1.060526	.0102173	6.10	0.000	1.040688 1.080742
charter		2.587793	1.797906	1.37	0.171	.6630481 10.09983
pyperstudlep		1.012983	.0059476	2.20	0.028	1.001393 1.024708
pyperstude~s		1.015417	.0037024	4.20	0.000	1.008187 1.0227
pyperstudw~e		1.017281	.0099654	1.75	0.080	.9979352 1.037001

pyperstudb~k		.9951643	.0103817	-0.46	0.642	.9750232	1.015722
pyperstudl~o		.9859538	.0099777	-1.40	0.162	.9665905	1.005705
pyperstudm~e		1.005794	.0044224	1.31	0.189	.9971633	1.014499
male		1.154829	.13064	1.27	0.203	.9251788	1.441484
age		.9709623	.0083758	-3.42	0.001	.9546841	.9875181
Asian		2.337599	2.013986	0.99	0.324	.4319236	12.65124
Black		1.273401	.2195018	1.40	0.161	.9083219	1.785215
Latino		1.826596	.3935074	2.80	0.005	1.197475	2.786241
Other		.4573351	.3041482	-1.18	0.239	.1242074	1.68392
AsianMale		.6338964	.6923063	-0.42	0.676	.074539	5.3908
BlackMale		1.606168	.4158056	1.83	0.067	.9670116	2.667783
LatinoMale		2.123541	.7306498	2.19	0.029	1.081898	4.168068
OtherMale		1.00599	.9449579	0.01	0.995	.1596002	6.340944

-----> region = 11

note: charter omitted because of collinearity  
Iteration 0: log likelihood = -2416.5808  
Iteration 1: log likelihood = -2276.098  
Iteration 2: log likelihood = -2273.5677  
Iteration 3: log likelihood = -2273.5623  
Iteration 4: log likelihood = -2273.5623

Multinomial logistic regression	Number of obs	=	2227
	LR chi2(36)	=	286.04
	Prob > chi2	=	0.0000
Log likelihood = -2273.5623	Pseudo R2	=	0.0592

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9972247	.0077332	-0.36	0.720	.9821826 1.012497
exper		1.001799	.0095726	0.19	0.851	.9832113 1.020737
charter		(omitted)				
pyperstudlep		.9962656	.0073222	-0.51	0.611	.9820172 1.010721
pyperstude~s		1.001281	.0038371	0.33	0.738	.9937883 1.00883
pyperstudw~e		.965598	.0130178	-2.60	0.009	.9404177 .9914525
pyperstudb~k		.9692344	.0140083	-2.16	0.031	.942164 .9970826
pyperstudl~o		.9709442	.01334	-2.15	0.032	.9451473 .9974453
pyperstudm~e		.9901108	.0042575	-2.31	0.021	.9818014 .9984905
male		1.121795	.1311625	0.98	0.326	.8920501 1.41071
age		.9768529	.0080559	-2.84	0.005	.9611905 .9927704
Asian		1.06193	.8229616	0.08	0.938	.2325103 4.850088
Black		1.587435	.3636921	2.02	0.044	1.013164 2.487207
Latino		2.508377	.8041199	2.87	0.004	1.338197 4.701817
Other		1.084003	.6465553	0.14	0.892	.3367675 3.489241
AsianMale		2.825848	3.966517	0.74	0.459	.1804493 44.25298
BlackMale		2.278457	.8446766	2.22	0.026	1.101744 4.711953
LatinoMale		.5881544	.2660401	-1.17	0.241	.2423634 1.427301
OtherMale		1.094166	.9954233	0.10	0.921	.1839492 6.508312
3						
tenure		1.008547	.0073163	1.17	0.241	.9943093 1.02299
exper		1.066079	.0112792	6.05	0.000	1.0442 1.088417
charter		(omitted)				
pyperstudlep		1.013026	.0083618	1.57	0.117	.9967688 1.029548
pyperstude~s		1.026011	.0042323	6.22	0.000	1.017749 1.034339
pyperstudw~e		1.007903	.0155169	0.51	0.609	.9779449 1.038779
pyperstudb~k		.9878938	.0163156	-0.74	0.461	.9564277 1.020395
pyperstudl~o		.9752014	.0155591	-1.57	0.116	.945178 1.006179
pyperstudm~e		.9983192	.0041143	-0.41	0.683	.9902878 1.006416
male		1.32442	.1613631	2.31	0.021	1.043081 1.681642

age		.9836805	.0095339	-1.70	0.090	.9651709	1.002545
Asian		.5744796	.6488897	-0.49	0.624	.0627799	5.256891
Black		1.671649	.4478986	1.92	0.055	.9887254	2.826275
Latino		3.342801	1.120505	3.60	0.000	1.732959	6.448115
Other		.5108185	.4155355	-0.83	0.409	.1037145	2.515901
AsianMale		5.647551	10.27539	0.95	0.341	.1596427	199.7888
BlackMale		1.460411	.6303771	0.88	0.380	.6266997	3.403226
LatinoMale		.5322453	.2614497	-1.28	0.199	.2032284	1.393925
OtherMale		1.960588	2.222153	0.59	0.553	.2126303	18.07789

-----> region = 12

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -860.19296  
Iteration 1: log likelihood = -811.47069  
Iteration 2: log likelihood = -809.63197  
Iteration 3: log likelihood = -809.43003  
Iteration 4: log likelihood = -809.39146  
Iteration 5: log likelihood = -809.3832  
Iteration 6: log likelihood = -809.38131  
Iteration 7: log likelihood = -809.38085  
Iteration 8: log likelihood = -809.38076  
Iteration 9: log likelihood = -809.38074

Multinomial logistic regression	Number of obs	=	811
	LR chi2(34)	=	101.62
	Prob > chi2	=	0.0000
Log likelihood = -809.38074	Pseudo R2	=	0.0591

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.012301	.0141197	0.88	0.381	.9850012 1.040356
exper		.9808282	.0178285	-1.06	0.287	.9465002 1.016401
charter		2.64e-07	.000727	-0.01	0.996	0 .
pyperstudlep		1.038691	.0246307	1.60	0.109	.9915206 1.088106
pyperstude~s		.9882414	.0073805	-1.58	0.113	.9738812 1.002813
pyperstudw~e		.9206883	.0516608	-1.47	0.141	.8248041 1.027719
pyperstudb~k		.9387769	.0565114	-1.05	0.294	.8343009 1.056336
pyperstudl~o		.925482	.051974	-1.38	0.168	.8290209 1.033167
pyperstudm~e		.997541	.0080906	-0.30	0.761	.981809 1.013525
male		.8349329	.1803927	-0.83	0.404	.5466935 1.275144
age		.9749011	.0152406	-1.63	0.104	.9454831 1.005234
Asian		(omitted)				
Black		.8808103	.305062	-0.37	0.714	.4467608 1.736559
Latino		2.755144	1.784467	1.56	0.118	.7741644 9.805176
Other		.2325132	.2736038	-1.24	0.215	.0231643 2.33387
AsianMale		(omitted)				
BlackMale		2.08889	1.367403	1.13	0.260	.5790458 7.535608
LatinoMale		.6842577	.6191215	-0.42	0.675	.1161572 4.030818
OtherMale		9.534445	17.93352	1.20	0.231	.2389207 380.4845
3						
tenure		.9980656	.0109769	-0.18	0.860	.9767815 1.019814
exper		1.043903	.0170101	2.64	0.008	1.011091 1.077781
charter		5.70e-07	.0013611	-0.01	0.995	0 .
pyperstudlep		1.059202	.0231763	2.63	0.009	1.014737 1.105615
pyperstude~s		1.009797	.0068522	1.44	0.151	.9964555 1.023317
pyperstudw~e		1.065302	.0619968	1.09	0.277	.9504641 1.194014
pyperstudb~k		1.055136	.06494	0.87	0.383	.9352336 1.190412
pyperstudl~o		1.032917	.060462	0.55	0.580	.9209588 1.158486



pyperstudm~e		1.003259	.0077848	0.42	0.675	.9881161	1.018633
male		1.128078	.212375	0.64	0.522	.779991	1.631506
age		.9837731	.0147222	-1.09	0.274	.9553371	1.013055
Asian		(omitted)					
Black		1.092668	.3824732	0.25	0.800	.5502208	2.169899
Latino		1.162861	.918287	0.19	0.848	.2473727	5.466431
Other		8.53e-07	.0007786	-0.02	0.988	0	.
AsianMale		(omitted)					
BlackMale		2.165457	1.309568	1.28	0.201	.6618852	7.08462
LatinoMale		.7073356	.821289	-0.30	0.766	.0726597	6.885851
OtherMale		1.085643	1897.392	0.00	1.000	0	.

-----> region = 13

Iteration 0: log likelihood = -1423.2109  
Iteration 1: log likelihood = -1353.8044  
Iteration 2: log likelihood = -1352.5434  
Iteration 3: log likelihood = -1352.3325  
Iteration 4: log likelihood = -1352.2984  
Iteration 5: log likelihood = -1352.2948  
Iteration 6: log likelihood = -1352.2941  
Iteration 7: log likelihood = -1352.2939  
Iteration 8: log likelihood = -1352.2939  
Iteration 9: log likelihood = -1352.2939

Multinomial logistic regression	Number of obs	=	1315
	LR chi2(38)	=	141.83
	Prob > chi2	=	0.0000
Log likelihood = -1352.2939	Pseudo R2	=	0.0498

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.029126	.0112325	2.63	0.009	1.007345 1.051379
exper		.9490886	.0119158	-4.16	0.000	.926019 .9727328
charter		2.31321	10871.57	0.00	1.000	0 .
pyperstudlep		.9923241	.0097641	-0.78	0.434	.9733701 1.011647
pyperstude~s		1.002178	.0072647	0.30	0.764	.98804 1.016518
pyperstudw~e		.9479712	.023874	-2.12	0.034	.9023151 .9959374
pyperstudb~k		.950402	.0269767	-1.79	0.073	.8989726 1.004774
pyperstudl~o		.9521384	.0245905	-1.90	0.058	.9051415 1.001576
pyperstudm~e		.9913002	.0064028	-1.35	0.176	.9788301 1.003929
male		.8777356	.1457738	-0.79	0.432	.6338649 1.215432
age		.9953219	.0108759	-0.43	0.668	.9742323 1.016868
Asian		.7309693	.9090815	-0.25	0.801	.0638688 8.365837
Black		1.256893	.4028444	0.71	0.476	.6706277 2.355675
Latino		1.174137	.2891424	0.65	0.514	.7246065 1.902546
Other		3.368367	4.226814	0.97	0.333	.2879282 39.40529
AsianMale		1.86e+07	5.40e+10	0.01	0.995	0 .
BlackMale		2.274906	1.213977	1.54	0.123	.7993366 6.474364
LatinoMale		1.891802	.744469	1.62	0.105	.8747975 4.091134
OtherMale		.3409501	.5038139	-0.73	0.467	.0188317 6.17293
3						
tenure		1.029758	.0102248	2.95	0.003	1.009911 1.049994
exper		.9953409	.0125542	-0.37	0.711	.9710368 1.020253
charter		1.36e+07	4.10e+10	0.01	0.996	0 .
pyperstudlep		.998331	.0094805	-0.18	0.860	.9799215 1.017086
pyperstude~s		1.033001	.0071853	4.67	0.000	1.019014 1.04718
pyperstudw~e		.9516817	.024994	-1.89	0.059	.9039339 1.001952
pyperstudb~k		.9168061	.0273519	-2.91	0.004	.8647347 .9720131
pyperstudl~o		.9337257	.0251636	-2.54	0.011	.8856858 .9843713

```

pyperstudm~e | .998268 .0060373 -0.29 0.774 .9865049 1.010171
  male | 1.040775 .1662993 0.25 0.802 .7609369 1.423526
  age | 1.001175 .0114789 0.10 0.918 .9789276 1.023928
  Asian | 1.173373 1.457569 0.13 0.898 .1028182 13.39067
  Black | .5339102 .2180624 -1.54 0.124 .2397817 1.188832
  Latino | .8297409 .2188285 -0.71 0.479 .494828 1.391332
  Other | 2.114278 2.681356 0.59 0.555 .1760583 25.39028
  AsianMale | 1.843573 8404.293 0.00 1.000 0 .
  BlackMale | 4.643549 2.835084 2.51 0.012 1.403315 15.36543
  LatinoMale | 2.002313 .8276323 1.68 0.093 .890631 4.501591
  OtherMale | 3.90e-07 .0003054 -0.02 0.985 0 .
-----

```

```

-----> region = 14

```

```

note: charter omitted because of collinearity
note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -274.60202
Iteration 1: log likelihood = -239.8176
Iteration 2: log likelihood = -235.6764
Iteration 3: log likelihood = -234.99504
Iteration 4: log likelihood = -234.94115
Iteration 5: log likelihood = -234.92991
Iteration 6: log likelihood = -234.92736
Iteration 7: log likelihood = -234.92673
Iteration 8: log likelihood = -234.9266
Iteration 9: log likelihood = -234.92658

```

```

Multinomial logistic regression          Number of obs   =      285
                                         LR chi2(30)      =      79.35
                                         Prob > chi2      =      0.0000
Log likelihood = -234.92658              Pseudo R2       =      0.1445

```

```

-----
      admin |      RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1          | (base outcome)
-----+-----
2          |
  tenure | 1.000687   .0341147     0.02   0.984   .9360088   1.069835
  exper | .9400442   .0392259    -1.48   0.138   .8662227   1.020157
  charter | (omitted)
pyperstudlep | .7874025   .1463387    -1.29   0.198   .5470173   1.133424
pyperstude~s | .9773142   .0176163    -1.27   0.203   .9433896   1.012459
pyperstudw~e | .8813598   .2467259    -0.45   0.652   .5091782   1.525586
pyperstudb~k | .9398044   .2903042    -0.20   0.841   .5129825   1.721759
pyperstudl~o | .894977    .2438941    -0.41   0.684   .5246228   1.52678
pyperstudm~e | .9931761   .0273604    -0.25   0.804   .9409726   1.048276
  male | 1.926957   .9337848     1.35   0.176   .7453977   4.981454
  age | 1.064338   .0401544     1.65   0.098   .9884765   1.146022
  Asian | (omitted)
  Black | 4.09e-06   .0029939    -0.02   0.986   0 .
  Latino | 3.904437   5.267705     1.01   0.313   .2774293   54.94959
  Other | 2.23e-06   .0042806    -0.01   0.995   0 .
  AsianMale | (omitted)
  BlackMale | 303766.5   5.13e+08     0.01   0.994   0 .
  LatinoMale | .7746799   1.191005    -0.17   0.868   .0380603   15.76784
  OtherMale | (omitted)
-----
3          |
  tenure | 1.010063   .0205707     0.49   0.623   .9705392   1.051196
  exper | 1.038119   .0302462     1.28   0.199   .9804982   1.099125
  charter | (omitted)
pyperstudlep | 1.037683   .0446003     0.86   0.389   .9538486   1.128885

```

pyperstude~s		1.037886	.0113729	3.39	0.001	1.015833	1.060418
pyperstudw~e		1.418688	.2595101	1.91	0.056	.9912495	2.030444
pyperstudb~k		1.35431	.2670713	1.54	0.124	.9201535	1.993315
pyperstudl~o		1.393268	.2508921	1.84	0.066	.9789378	1.982962
pyperstudm~e		.9957021	.0168807	-0.25	0.799	.9631601	1.029344
male		2.443693	.7373341	2.96	0.003	1.352739	4.414477
age		.9914219	.0262335	-0.33	0.745	.9413157	1.044195
Asian		(omitted)					
Black		3.176156	3.462244	1.06	0.289	.3749984	26.90136
Latino		1.392632	1.426963	0.32	0.747	.1869177	10.37581
Other		9.50e-07	.0009594	-0.01	0.989	0	.
AsianMale		(omitted)					
BlackMale		202361.9	1.47e+08	0.02	0.987	0	.
LatinoMale		.4738362	.6546807	-0.54	0.589	.0315905	7.107218
OtherMale		(omitted)					

-----> region = 15

note: charter omitted because of collinearity  
note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -284.28468  
Iteration 1: log likelihood = -251.01136  
Iteration 2: log likelihood = -246.54867  
Iteration 3: log likelihood = -246.36963  
Iteration 4: log likelihood = -246.34192  
Iteration 5: log likelihood = -246.3375  
Iteration 6: log likelihood = -246.33643  
Iteration 7: log likelihood = -246.33621  
Iteration 8: log likelihood = -246.33616  
Iteration 9: log likelihood = -246.33615

Multinomial logistic regression	Number of obs	=	289
	LR chi2(30)	=	75.90
	Prob > chi2	=	0.0000
Log likelihood = -246.33615	Pseudo R2	=	0.1335

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9544413	.024866	-1.79	0.073	.9069282 1.004444
exper		.9765627	.0412867	-0.56	0.575	.8989041 1.06093
charter		(omitted)				
pyperstudlep		1.052845	.032421	1.67	0.094	.991181 1.118346
pyperstude~s		.9751501	.0160356	-1.53	0.126	.944222 1.007091
pyperstudw~e		.7443762	.2461221	-0.89	0.372	.3893583 1.4231
pyperstudb~k		.8747924	.308837	-0.38	0.705	.4379218 1.747485
pyperstudl~o		.7598737	.249751	-0.84	0.403	.3990004 1.447137
pyperstudm~e		.9708088	.02627	-1.09	0.274	.920662 1.023687
male		4.429719	2.635105	2.50	0.012	1.380455 14.21445
age		1.083727	.0458323	1.90	0.057	.9975191 1.177384
Asian		(omitted)				
Black		3.02e-06	.0072687	-0.01	0.996	0 .
Latino		15.22664	11.38327	3.64	0.000	3.517669 65.91029
Other		2.788557	8884.695	0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		663824.3	2.65e+09	0.00	0.997	0 .
LatinoMale		.1228545	.1112694	-2.32	0.021	.0208189 .7249782
OtherMale		(omitted)				
3						

tenure		.9417605	.0171891	-3.29	0.001	.9086661	.9760603
exper		1.048841	.0304205	1.64	0.100	.9908812	1.110192
charter		(omitted)					
pyperstudlep		1.043712	.0265764	1.68	0.093	.9929016	1.097122
pyperstude~s		1.023585	.0104199	2.29	0.022	1.003365	1.044213
pyperstudw~e		1.153592	.2452536	0.67	0.502	.7604786	1.749918
pyperstudb~k		1.24674	.2877375	0.96	0.339	.7930958	1.959866
pyperstudl~o		1.129782	.2393673	0.58	0.565	.7458476	1.711351
pyperstudm~e		.9785344	.0155803	-1.36	0.173	.9484692	1.009553
male		1.534186	.4828751	1.36	0.174	.8278807	2.843074
age		1.001353	.0281214	0.05	0.962	.9477258	1.058015
Asian		(omitted)					
Black		4.50e-06	.0064042	-0.01	0.993	0	.
Latino		1.140762	.7448942	0.20	0.840	.3172324	4.102157
Other		5032678	7.37e+09	0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		8.62e+11	1.76e+15	0.01	0.989	0	.
LatinoMale		1.013851	.8248292	0.02	0.987	.2058118	4.99434
OtherMale		(omitted)					

-----> region = 16

note: charter omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -477.06084  
Iteration 1: log likelihood = -431.73602  
Iteration 2: log likelihood = -429.11879  
Iteration 3: log likelihood = -428.85606  
Iteration 4: log likelihood = -428.79852  
Iteration 5: log likelihood = -428.78586  
Iteration 6: log likelihood = -428.7833  
Iteration 7: log likelihood = -428.78273  
Iteration 8: log likelihood = -428.78259  
Iteration 9: log likelihood = -428.78256

Multinomial logistic regression	Number of obs	=	452
	LR chi2(34)	=	96.56
	Prob > chi2	=	0.0000
Log likelihood = -428.78256	Pseudo R2	=	0.1012

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.060916	.0248174	2.53	0.011	1.013373 1.11069
exper		.9247333	.0275867	-2.62	0.009	.8722148 .9804141
charter		(omitted)				
pyperstudlep		.9952976	.0236822	-0.20	0.843	.949947 1.042813
pyperstude~s		.9740752	.0099212	-2.58	0.010	.9548227 .9937158
pyperstudw~e		.9118385	.0462655	-1.82	0.069	.8255227 1.007179
pyperstudb~k		.9511064	.054353	-0.88	0.380	.8503259 1.063831
pyperstudl~o		.9291471	.0478491	-1.43	0.154	.8399422 1.027826
pyperstudm~e		1.036474	.0153907	2.41	0.016	1.006744 1.067083
male		2.091589	.6446202	2.39	0.017	1.143244 3.826604
age		1.003753	.0251959	0.15	0.881	.9555653 1.054371
Asian		9346995	2.67e+10	0.01	0.996	0
Black		.2599164	.335061	-1.05	0.296	.020775 3.25182
Latino		12.40364	9.466223	3.30	0.001	2.779254 55.35668
Other		1.617405	2.505793	0.31	0.756	.077639 33.69439
AsianMale		(omitted)				
BlackMale		7.56e-07	.0005773	-0.02	0.985	0
LatinoMale		.1723272	.2184502	-1.39	0.165	.0143658 2.067175
OtherMale		1190235	1.01e+09	0.02	0.987	0

-----+-----						
3						
tenure		1.026204	.0158397	1.68	0.094	.995624 1.057724
exper		1.02238	.0238007	0.95	0.342	.97678 1.07011
charter		(omitted)				
pyperstudlep		1.037057	.0185508	2.03	0.042	1.001328 1.074061
pyperstude~s		1.00144	.0075344	0.19	0.848	.986781 1.016316
pyperstudw~e		.9592939	.0429441	-0.93	0.353	.8787118 1.047266
pyperstudb~k		.9602986	.0476187	-0.82	0.414	.8713595 1.058316
pyperstudl~o		.9510476	.0425805	-1.12	0.262	.8711482 1.038275
pyperstudm~e		1.016864	.0143703	1.18	0.237	.9890855 1.045423
male		1.628759	.3874254	2.05	0.040	1.021846 2.596141
age		1.000871	.0217274	0.04	0.968	.9591789 1.044374
Asian		1.41273	5993.725	0.00	1.000	0 .
Black		6.73e-07	.0003911	-0.02	0.980	0 .
Latino		1.412578	1.269985	0.38	0.701	.2425135 8.2279
Other		.5452957	.6940719	-0.48	0.634	.0449979 6.608028
AsianMale		(omitted)				
BlackMale		231793.1	1.35e+08	0.02	0.983	0 .
LatinoMale		1.26453	1.643266	0.18	0.857	.0990345 16.14626
OtherMale		511040.4	4.36e+08	0.02	0.988	0 .
-----+-----						

-----> region = 17

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -451.15759  
Iteration 1: log likelihood = -406.08772  
Iteration 2: log likelihood = -403.87854  
Iteration 3: log likelihood = -403.22481  
Iteration 4: log likelihood = -403.15988  
Iteration 5: log likelihood = -403.14821  
Iteration 6: log likelihood = -403.14565  
Iteration 7: log likelihood = -403.14502  
Iteration 8: log likelihood = -403.14489  
Iteration 9: log likelihood = -403.14486

Multinomial logistic regression	Number of obs	=	453
	LR chi2(34)	=	96.03
	Prob > chi2	=	0.0000
Log likelihood = -403.14486	Pseudo R2	=	0.1064

-----+-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
1		(base outcome)				
-----+-----						
2						
tenure		1.00919	.0208042	0.44	0.657	.9692274 1.0508
exper		1.013815	.0315918	0.44	0.660	.9537489 1.077664
charter		1.571143	2723.354	0.00	1.000	0 .
pyperstudlep		.9831877	.0547097	-0.30	0.761	.8815991 1.096483
pyperstude~s		.9697208	.0133558	-2.23	0.026	.9438941 .9962541
pyperstudw~e		.9375342	.1407604	-0.43	0.667	.6985357 1.258304
pyperstudb~k		.9754431	.147232	-0.16	0.869	.7256426 1.311237
pyperstudl~o		.9594959	.1435167	-0.28	0.782	.7156894 1.286358
pyperstudm~e		.9930346	.011323	-0.61	0.540	.971088 1.015477
male		3.345473	1.177164	3.43	0.001	1.678599 6.667575
age		1.009737	.0295452	0.33	0.741	.9534588 1.069338
Asian		5.24e-06	.007077	-0.01	0.993	0 .
Black		2.754153	2.788212	1.00	0.317	.3786695 20.03161
Latino		1.299777	1.452969	0.23	0.815	.1453264 11.625
Other		2023933	1.36e+09	0.02	0.983	0 .
AsianMale		(omitted)				
BlackMale		1414407	1.55e+09	0.01	0.990	0 .

LatinoMale		1.215204	1.608915	0.15	0.883	.0907128	16.27908
OtherMale		(omitted)					
-----							
3							
tenure		1.01793	.0159121	1.14	0.256	.9872159	1.0496
exper		1.033256	.0244383	1.38	0.167	.9864508	1.082282
charter		626685	5.18e+08	0.02	0.987	0	.
pyperstudlep		1.066198	.0295572	2.31	0.021	1.009813	1.125732
pyperstude~s		1.010457	.0099839	1.05	0.292	.9910776	1.030216
pyperstudw~e		1.071614	.1246919	0.59	0.552	.8530874	1.346119
pyperstudb~k		1.062202	.1246493	0.51	0.607	.8439543	1.33689
pyperstudl~o		1.069879	.124615	0.58	0.562	.8515111	1.344247
pyperstudm~e		.9873103	.0095844	-1.32	0.188	.9687028	1.006275
male		2.800973	.6731242	4.29	0.000	1.748839	4.48609
age		.9844896	.0213121	-0.72	0.470	.9435924	1.027159
Asian		3.33e-06	.0026868	-0.02	0.988	0	.
Black		.4631514	.5704038	-0.62	0.532	.0414389	5.176523
Latino		2.237004	1.307959	1.38	0.169	.7111737	7.03652
Other		292941.6	1.96e+08	0.02	0.985	0	.
AsianMale		(omitted)					
BlackMale		3690181	4.05e+09	0.01	0.989	0	.
LatinoMale		.5768986	.4654019	-0.68	0.495	.1186886	2.804078
OtherMale		(omitted)					
-----							

-----> region = 18

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -426.48913  
Iteration 1: log likelihood = -397.11821  
Iteration 2: log likelihood = -396.04368  
Iteration 3: log likelihood = -395.89892  
Iteration 4: log likelihood = -395.86341  
Iteration 5: log likelihood = -395.85687  
Iteration 6: log likelihood = -395.85576  
Iteration 7: log likelihood = -395.8555  
Iteration 8: log likelihood = -395.85544  
Iteration 9: log likelihood = -395.85542

Multinomial logistic regression	Number of obs	=	403
	LR chi2(34)	=	61.27
	Prob > chi2	=	0.0028
Log likelihood = -395.85542	Pseudo R2	=	0.0718

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.016153	.0222255	0.73	0.464	.9735129 1.060662
exper		.954085	.0256365	-1.75	0.080	.9051386 1.005678
charter		2.78e+07	5.20e+10	0.01	0.993	0 .
pyperstudlep		1.019099	.0175268	1.10	0.271	.9853194 1.054036
pyperstude~s		.9877108	.0128457	-0.95	0.342	.9628519 1.013212
pyperstudw~e		.9786202	.0685902	-0.31	0.758	.853011 1.122726
pyperstudb~k		1.015325	.077491	0.20	0.842	.8742589 1.179152
pyperstudl~o		.9836018	.0683505	-0.24	0.812	.8583597 1.127118
pyperstudm~e		1.012311	.0128785	0.96	0.336	.9873813 1.037869
male		1.771536	.5952653	1.70	0.089	.9169217 3.422691
age		.9860837	.0229751	-0.60	0.548	.942066 1.032158
Asian		(omitted)				
Black		5.747627	7.331797	1.37	0.170	.4717127 70.03249
Latino		2.495924	1.193267	1.91	0.056	.9778757 6.37058
Other		2.234011	3.340798	0.54	0.591	.1191726 41.87879

AsianMale		(omitted)					
BlackMale		.0816496	.1695476	-1.21	0.228	.0013945	4.780762
LatinoMale		.4544373	.2897369	-1.24	0.216	.130248	1.585539
OtherMale		2.91e-08	.0000877	-0.01	0.995	0	.
-----							
3							
tenure		1.022248	.0170956	1.32	0.188	.9892849	1.05631
exper		1.054496	.0274917	2.04	0.042	1.001967	1.109779
charter		2744644	5.13e+09	0.01	0.994	0	.
pyperstudlep		.9998148	.0145786	-0.01	0.990	.9716456	1.028801
pyperstude~s		1.023989	.0112859	2.15	0.031	1.002106	1.046349
pyperstudw~e		1.001802	.085391	0.02	0.983	.847672	1.183958
pyperstudb~k		.9876234	.0892494	-0.14	0.890	.8273136	1.178997
pyperstudl~o		.9964754	.0842176	-0.04	0.967	.8443585	1.175997
pyperstudm~e		1.002247	.012075	0.19	0.852	.9788578	1.026196
male		1.773313	.4958275	2.05	0.040	1.025144	3.067508
age		.9367191	.0228808	-2.68	0.007	.8929302	.9826554
Asian		(omitted)					
Black		4.927714	5.760879	1.36	0.172	.4983391	48.72658
Latino		1.198746	.5345701	0.41	0.684	.5001976	2.87285
Other		3.80e-07	.0008302	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		.4469593	.759656	-0.47	0.636	.0159793	12.50195
LatinoMale		.5274215	.3137461	-1.08	0.282	.1643635	1.692428
OtherMale		.486824	1484.397	-0.00	1.000	0	.
-----							

-----> region = 19

Iteration 0: log likelihood = -804.20395  
Iteration 1: log likelihood = -731.3405  
Iteration 2: log likelihood = -729.05119  
Iteration 3: log likelihood = -728.94158  
Iteration 4: log likelihood = -728.92189  
Iteration 5: log likelihood = -728.91997  
Iteration 6: log likelihood = -728.91955  
Iteration 7: log likelihood = -728.91945  
Iteration 8: log likelihood = -728.91944

Multinomial logistic regression	Number of obs	=	751
	LR chi2(38)	=	150.57
	Prob > chi2	=	0.0000
Log likelihood = -728.91944	Pseudo R2	=	0.0936

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.02107	.0137872	1.54	0.123	.9944022 1.048453
exper		1.055271	.0200683	2.83	0.005	1.016662 1.095346
charter		1.45e-06	.0016193	-0.01	0.990	0
pyperstudlep		1.01256	.0059783	2.11	0.035	1.000911 1.024346
pyperstude~s		.9942363	.0089517	-0.64	0.521	.9768452 1.011937
pyperstudw~e		1.004895	.1050154	0.05	0.963	.8187789 1.233316
pyperstudb~k		1.073139	.1173365	0.65	0.519	.8661359 1.329615
pyperstudl~o		1.00671	.0984166	0.07	0.945	.8311708 1.219322
pyperstudm~e		1.000147	.0076794	0.02	0.985	.9852086 1.015312
male		1.702363	.4943198	1.83	0.067	.963577 3.007585
age		.9635152	.0153762	-2.33	0.020	.9338448 .9941283
Asian		.9081549	4401.819	-0.00	1.000	0
Black		.7639696	.5448952	-0.38	0.706	.1887832 3.091639
Latino		1.674345	.4267478	2.02	0.043	1.016004 2.759272
Other		9532549	2.02e+10	0.01	0.994	0
AsianMale		.3839457	2631.825	-0.00	1.000	0

BlackMale		.5487833	.6083395	-0.54	0.588	.0624919	4.819237
LatinoMale		.8009792	.2936397	-0.61	0.545	.3904541	1.643132
OtherMale		1.58e-13	4.48e-10	-0.01	0.992	0	.
-----							
3							
tenure		1.043825	.0152094	2.94	0.003	1.014436	1.074064
exper		1.101223	.0246084	4.31	0.000	1.054033	1.150526
charter		4.653046	7.377172	0.97	0.332	.2080654	104.0578
pyperstudlep		1.023579	.0068904	3.46	0.001	1.010163	1.037173
pyperstude~s		1.011464	.0105756	1.09	0.276	.9909467	1.032405
pyperstudw~e		.9117788	.0995009	-0.85	0.397	.7362058	1.129223
pyperstudb~k		.9609497	.1082239	-0.35	0.724	.7706139	1.198297
pyperstudl~o		.9001028	.0911126	-1.04	0.298	.7381244	1.097627
pyperstudm~e		1.004528	.0081413	0.56	0.577	.9886978	1.020613
male		1.273506	.4240657	0.73	0.468	.6630687	2.445925
age		.9777948	.0188836	-1.16	0.245	.9414753	1.015515
Asian		.9977466	3.16e+10	0.01	0.996	0	.
Black		.7121665	.6031655	-0.40	0.689	.1354132	3.745432
Latino		1.373764	.3855139	1.13	0.258	.7925794	2.381121
Other		2.199976	7774.514	0.00	1.000	0	.
AsianMale		.2092585	937.3837	-0.00	1.000	0	.
BlackMale		2.904968	3.198777	0.97	0.333	.3356211	25.14395
LatinoMale		.7291396	.313114	-0.74	0.462	.3142532	1.691771
OtherMale		1.09e-06	.0045216	-0.00	0.997	0	.
-----							

-----> region = 20

Iteration 0: log likelihood = -2163.5084  
Iteration 1: log likelihood = -2036.3755  
Iteration 2: log likelihood = -2029.5237  
Iteration 3: log likelihood = -2028.9128  
Iteration 4: log likelihood = -2028.7837  
Iteration 5: log likelihood = -2028.7558  
Iteration 6: log likelihood = -2028.7512  
Iteration 7: log likelihood = -2028.7501  
Iteration 8: log likelihood = -2028.7499  
Iteration 9: log likelihood = -2028.7499  
Iteration 10: log likelihood = -2028.7498

Multinomial logistic regression	Number of obs	=	2149
	LR chi2(38)	=	269.52
	Prob > chi2	=	0.0000
Log likelihood = -2028.7498	Pseudo R2	=	0.0623

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.000277	.0081247	0.03	0.973	.984479 1.016329
exper		1.000683	.0104556	0.07	0.948	.9803992 1.021387
charter		3.41e-07	.0004645	-0.01	0.991	0 .
pyperstudlep		.9860863	.007419	-1.86	0.063	.971652 1.000735
pyperstude~s		.9988632	.0047912	-0.24	0.813	.9895167 1.008298
pyperstudw~e		.9858098	.0391697	-0.36	0.719	.9119519 1.065649
pyperstudb~k		.9845658	.040982	-0.37	0.709	.9074317 1.068257
pyperstudl~o		.9867348	.0391212	-0.34	0.736	.912962 1.066469
pyperstudm~e		1.000411	.0042215	0.10	0.922	.9921715 1.00872
male		1.444028	.2077116	2.55	0.011	1.089274 1.914318
age		.9719119	.0086454	-3.20	0.001	.955114 .9890052
Asian		1.92e-07	.000243	-0.01	0.990	0 .
Black		1.446614	.432372	1.24	0.217	.8052685 2.59875
Latino		1.257101	.2013236	1.43	0.153	.9184403 1.720637
Other		1.936528	1.504969	0.85	0.395	.422196 8.882465



AsianMale		1.41e+14	6.39e+17	0.01	0.994	0	.
BlackMale		1.099034	.5097505	0.20	0.839	.4428023	2.727801
LatinoMale		1.407449	.3313934	1.45	0.147	.8871795	2.232821
OtherMale		1.67e-07	.0002199	-0.01	0.991	0	.
-----							
3							
tenure		1.021217	.0081054	2.65	0.008	1.005454	1.037228
exper		1.092765	.0145981	6.64	0.000	1.064525	1.121755
charter		3.768346	3.274579	1.53	0.127	.6862499	20.6928
pyperstudlep		1.02437	.0067828	3.64	0.000	1.011162	1.03775
pyperstude~s		1.02253	.0056635	4.02	0.000	1.01149	1.033691
pyperstudw~e		1.10196	.0524119	2.04	0.041	1.003877	1.209626
pyperstudb~k		1.066589	.0530851	1.30	0.195	.9674578	1.175878
pyperstudl~o		1.067203	.0509734	1.36	0.173	.9718311	1.171935
pyperstudm~e		1.009367	.004405	2.14	0.033	1.00077	1.018038
male		1.423562	.2207173	2.28	0.023	1.050512	1.929086
age		.9412916	.0114277	-4.98	0.000	.919158	.9639581
Asian		2.95e-07	.0003854	-0.01	0.991	0	.
Black		1.274556	.3897995	0.79	0.428	.6998979	2.321044
Latino		.9140471	.1547578	-0.53	0.596	.6559205	1.273755
Other		6.60e-07	.0007736	-0.01	0.990	0	.
AsianMale		6867328	5.69e+10	0.00	0.998	0	.
BlackMale		.9398883	.4865199	-0.12	0.905	.3407719	2.592321
LatinoMale		1.068903	.2801264	0.25	0.799	.6395351	1.786536
OtherMale		4145138	4.86e+09	0.01	0.990	0	.

by pyschooltype:

```
. bysort pyschooltype : mlogit admin tenure exper charter pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile m
> ale age Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if
validcert2002a==1, rrr iter(20)
```

-----> pyschooltype =

note: OtherMale omitted because of collinearity

```
Iteration 0: log likelihood = -416.12364
Iteration 1: log likelihood = -361.70205
Iteration 2: log likelihood = -359.23751
Iteration 3: log likelihood = -358.99212
Iteration 4: log likelihood = -358.93735
Iteration 5: log likelihood = -358.92496
Iteration 6: log likelihood = -358.92225
Iteration 7: log likelihood = -358.92185
Iteration 8: log likelihood = -358.9218
Iteration 9: log likelihood = -358.92179
```

Multinomial logistic regression	Number of obs	=	388
	LR chi2(36)	=	114.40
	Prob > chi2	=	0.0000
Log likelihood = -358.92179	Pseudo R2	=	0.1375

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.040388	.0231976	1.78	0.076	.9959012	1.086863
	exper	1.039078	.0295033	1.35	0.177	.9828325	1.098543
	charter	1.06e-06	.0011161	-0.01	0.990	0	.
	pyperstudlep	1.002098	.0137416	0.15	0.879	.9755236	1.029396
	pyperstude~s	.9892429	.010686	-1.00	0.317	.9685188	1.01041
	pyperstudw~e	1.019065	.0249492	0.77	0.440	.9713206	1.069157
	pyperstudb~k	1.03012	.0294374	1.04	0.299	.9740094	1.089462
	pyperstudl~o	1.028672	.0275033	1.06	0.290	.9761548	1.084015
	pyperstudm~e	.9797127	.0139037	-1.44	0.149	.9528375	1.007346

male		2.379315	.7851888	2.63	0.009	1.246092	4.543115
age		.9511188	.0229014	-2.08	0.037	.9072756	.9970806
Asian		5.185499	6.856702	1.24	0.213	.3883831	69.23421
Black		.6754868	.3994733	-0.66	0.507	.2119475	2.152808
Latino		1.205835	.5908772	0.38	0.702	.4615152	3.150572
Other		3.61e-07	.0007373	-0.01	0.994	0	.
AsianMale		3.25e-08	.0000939	-0.01	0.995	0	.
BlackMale		1.837092	2.05664	0.54	0.587	.2047404	16.48383
LatinoMale		1.100758	.7701821	0.14	0.891	.279332	4.337739
OtherMale		(omitted)					
-----							
3							
tenure		1.081181	.0243865	3.46	0.001	1.034425	1.13005
exper		1.02823	.0295041	0.97	0.332	.9719992	1.087714
charter		8.009124	8.611515	1.94	0.053	.9735584	65.88827
pypersstudlep		1.003168	.0145916	0.22	0.828	.9749732	1.032179
pypersstude~s		1.023058	.011516	2.03	0.043	1.000734	1.045879
pypersstudw~e		1.103893	.0458953	2.38	0.017	1.017508	1.197613
pypersstudb~k		1.081344	.0489777	1.73	0.084	.9894871	1.181728
pypersstudl~o		1.076624	.0470108	1.69	0.091	.9883175	1.172821
pypersstudm~e		.9759565	.0153225	-1.55	0.121	.9463822	1.006455
male		1.577289	.5606137	1.28	0.200	.7859048	3.165576
age		1.017082	.0240412	0.72	0.474	.971037	1.065311
Asian		3.02e-06	.0041467	-0.01	0.993	0	.
Black		.5014562	.3325928	-1.04	0.298	.1366691	1.839907
Latino		.8481147	.4712909	-0.30	0.767	.2853963	2.52035
Other		3.91e-07	.0007561	-0.01	0.994	0	.
AsianMale		.0442196	139.3584	-0.00	0.999	0	.
BlackMale		13.74023	16.3738	2.20	0.028	1.329364	142.0183
LatinoMale		1.362716	1.101561	0.38	0.702	.2794672	6.644774
OtherMale		(omitted)					
-----							

-----> pyschooltype = B

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -629.13922
Iteration 1: log likelihood = -592.24503
Iteration 2: log likelihood = -587.2169
Iteration 3: log likelihood = -586.8926
Iteration 4: log likelihood = -586.82625
Iteration 5: log likelihood = -586.81234
Iteration 6: log likelihood = -586.80924
Iteration 7: log likelihood = -586.80848
Iteration 8: log likelihood = -586.80832
Iteration 9: log likelihood = -586.80829
Iteration 10: log likelihood = -586.80828 (not concave)
Iteration 11: log likelihood = -586.80828 (not concave)
Iteration 12: log likelihood = -586.80828 (not concave)
Iteration 13: log likelihood = -586.80828 (not concave)
Iteration 14: log likelihood = -586.80828 (not concave)
Iteration 15: log likelihood = -586.80828 (not concave)
Iteration 16: log likelihood = -586.80828 (not concave)
Iteration 17: log likelihood = -586.80828 (not concave)
Iteration 18: log likelihood = -586.80828 (not concave)
Iteration 19: log likelihood = -586.80828 (not concave)
Iteration 20: log likelihood = -586.80828 (not concave)
convergence not achieved

```

Multinomial logistic regression	Number of obs	=	663
	LR chi2(35)	=	84.66
	Prob > chi2	=	0.0000
Log likelihood = -586.80828	Pseudo R2	=	0.0673

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-------	--	-----	-----------	---	------	----------------------

1		(base outcome)					
2							
	tenure	1.046406	.0215824	2.20	0.028	1.004949	1.089573
	exper	.943274	.0235501	-2.34	0.019	.8982278	.9905793
	charter	1.99e-12	8.12e-07	-0.00	1.000	0	.
	pyperstudlep	1.01932	.010378	1.88	0.060	.9991807	1.039864
	pyperstude~s	.9901945	.0072354	-1.35	0.177	.9761144	1.004478
	pyperstudw~e	.9569748	.0366494	-1.15	0.251	.887773	1.031571
	pyperstudb~k	.97603	.0392715	-0.60	0.547	.9020159	1.056117
	pyperstudl~o	.9667683	.0374109	-0.87	0.382	.896156	1.042945
	pyperstudm~e	1.002037	.0045967	0.44	0.657	.9930677	1.011087
	male	1.232751	.426709	0.60	0.546	.6255143	2.429481
	age	1.001546	.0207811	0.07	0.941	.961633	1.043116
	Asian	2.00e-12	2.24e-06	-0.00	1.000	0	.
	Black	1.671833	.870933	0.99	0.324	.6022322	4.641112
	Latino	1.300329	.878686	0.39	0.698	.3458289	4.889284
	Other	3.89e+09	5.03e+13	0.00	0.999	0	.
	AsianMale	(omitted)					
	BlackMale	1.33011	.9419874	0.40	0.687	.3319439	5.329797
	LatinoMale	2.311566	1.803472	1.07	0.283	.5009602	10.66619
	OtherMale	1.42e-09	.0000184	-0.00	0.999	0	.
3							
	tenure	1.007158	.0119415	0.60	0.547	.9840231	1.030837
	exper	1.030118	.0171576	1.78	0.075	.9970325	1.064301
	charter	1.847801	.956963	1.19	0.236	.6696132	5.099016
	pyperstudlep	.9973836	.0097008	-0.27	0.788	.9785504	1.016579
	pyperstude~s	1.003997	.0051728	0.77	0.439	.9939094	1.014187
	pyperstudw~e	1.031936	.0354965	0.91	0.361	.9646571	1.103906
	pyperstudb~k	1.031635	.0368947	0.87	0.384	.9617992	1.106542
	pyperstudl~o	1.031169	.0357561	0.89	0.376	.9634163	1.103685
	pyperstudm~e	1.004196	.0032583	1.29	0.197	.9978297	1.010602
	male	1.217263	.2457541	0.97	0.330	.8194731	1.808149
	age	.969302	.0153389	-1.97	0.049	.9396997	.9998368
	Asian	.8924037	1.126281	-0.09	0.928	.0752108	10.5887
	Black	.7331307	.3237736	-0.70	0.482	.3085063	1.742203
	Latino	.8903624	.4537904	-0.23	0.820	.3278956	2.417676
	Other	1.013945	.945594	0.01	0.988	.1630021	6.307186
	AsianMale	(omitted)					
	BlackMale	1.740287	1.050364	0.92	0.359	.533177	5.680286
	LatinoMale	1.502258	.9318474	0.66	0.512	.4453958	5.066908
	OtherMale	2.714069	.	.	.	.	.

Warning: convergence not achieved

-----> pyschooltype = E

Iteration 0: log likelihood = -9915.2681  
Iteration 1: log likelihood = -9307.8832  
Iteration 2: log likelihood = -9302.1873  
Iteration 3: log likelihood = -9302.1836  
Iteration 4: log likelihood = -9302.1836

Multinomial logistic regression	Number of obs	=	9215
	LR chi2(38)	=	1226.17
	Prob > chi2	=	0.0000
Log likelihood = -9302.1836	Pseudo R2	=	0.0618

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					

tenure		1.002729	.0044052	0.62	0.535	.9941316	1.0114
exper		1.006348	.0058745	1.08	0.278	.9949	1.017928
charter		.5079197	.4386827	-0.78	0.433	.0934598	2.760357
pyperstudlep		1.011915	.0020876	5.74	0.000	1.007832	1.016015
pyperstude~s		.9819015	.0022774	-7.87	0.000	.977448	.9863753
pyperstudw~e		.9948658	.0065373	-0.78	0.433	.9821351	1.007762
pyperstudb~k		1.001603	.0071927	0.22	0.824	.9876039	1.0158
pyperstudl~o		1.000345	.006865	0.05	0.960	.9869802	1.013891
pyperstudm~e		1.013823	.0037897	3.67	0.000	1.006422	1.021277
male		1.769211	.1601698	6.30	0.000	1.481559	2.112713
age		.971009	.0045942	-6.22	0.000	.9620461	.9800553
Asian		.9727791	.447222	-0.06	0.952	.3950808	2.395204
Black		1.289281	.1368353	2.39	0.017	1.047146	1.587407
Latino		1.090785	.0977636	0.97	0.332	.915058	1.300258
Other		.6453753	.2430886	-1.16	0.245	.3084585	1.350293
AsianMale		2.071521	1.743099	0.87	0.387	.3981425	10.77805
BlackMale		.8207584	.1726332	-0.94	0.348	.5434742	1.239515
LatinoMale		.6910558	.1117733	-2.28	0.022	.5033122	.9488309
OtherMale		.5466787	.4244764	-0.78	0.437	.1193451	2.504146
-----							
3							
tenure		1.018288	.0036768	5.02	0.000	1.011107	1.02552
exper		1.064161	.0056614	11.69	0.000	1.053122	1.075315
charter		3.30851	2.017365	1.96	0.050	1.00141	10.93082
pyperstudlep		1.001527	.0018904	0.81	0.419	.9978285	1.005239
pyperstude~s		1.002443	.002074	1.18	0.238	.9983866	1.006516
pyperstudw~e		1.010377	.0067041	1.56	0.120	.9973225	1.023603
pyperstudb~k		.9998559	.0072106	-0.02	0.984	.9858227	1.014089
pyperstudl~o		1.000611	.0069171	0.09	0.930	.987145	1.01426
pyperstudm~e		1.002534	.0035863	0.71	0.479	.9955294	1.009588
male		2.923842	.2276267	13.78	0.000	2.510073	3.40582
age		.9824739	.0044549	-3.90	0.000	.9737812	.9912442
Asian		.6758191	.3673565	-0.72	0.471	.2328846	1.961192
Black		1.00133	.1038202	0.01	0.990	.8171892	1.226964
Latino		1.184218	.0983511	2.04	0.042	1.006325	1.393559
Other		.5870847	.2103736	-1.49	0.137	.2908604	1.184996
AsianMale		.8632089	.8407872	-0.15	0.880	.1279453	5.823816
BlackMale		.5858149	.1175184	-2.67	0.008	.3953699	.8679949
LatinoMale		.5362033	.0777074	-4.30	0.000	.4036195	.7123393
OtherMale		.2632876	.2058887	-1.71	0.088	.0568588	1.219166
-----							

-----> pyschooltype = M

Iteration 0: log likelihood = -5525.7566  
Iteration 1: log likelihood = -5219.576  
Iteration 2: log likelihood = -5214.0921  
Iteration 3: log likelihood = -5213.8313  
Iteration 4: log likelihood = -5213.7702  
Iteration 5: log likelihood = -5213.7584  
Iteration 6: log likelihood = -5213.7564  
Iteration 7: log likelihood = -5213.756  
Iteration 8: log likelihood = -5213.7559  
Iteration 9: log likelihood = -5213.7558

Multinomial logistic regression	Number of obs	=	5232
	LR chi2(38)	=	624.00
	Prob > chi2	=	0.0000
Log likelihood = -5213.7558	Pseudo R2	=	0.0565

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						

tenure		1.007826	.0050848	1.55	0.122	.9979089	1.017841
exper		.9978592	.0064219	-0.33	0.739	.9853517	1.010526
charter		4.01e-07	.0012115	-0.00	0.996	0	.
pyperstudlep		1.010875	.0039096	2.80	0.005	1.003241	1.018567
pyperstude~s		1.001746	.0028017	0.62	0.533	.9962696	1.007252
pyperstudw~e		.9799973	.0087527	-2.26	0.024	.9629916	.9973034
pyperstudb~k		.9834128	.0096212	-1.71	0.087	.9647352	1.002452
pyperstudl~o		.9758559	.0092321	-2.58	0.010	.957928	.9941193
pyperstudm~e		1.001249	.0037818	0.33	0.741	.9938645	1.008689
male		1.496611	.1236718	4.88	0.000	1.272829	1.759736
age		.959904	.0052195	-7.53	0.000	.9497283	.9701888
Asian		.4958067	.3466623	-1.00	0.316	.1259395	1.951923
Black		1.039604	.1359342	0.30	0.766	.8045784	1.343283
Latino		1.929151	.2543876	4.98	0.000	1.489781	2.498101
Other		.6214677	.3021381	-0.98	0.328	.2396575	1.611558
AsianMale		7.423556	8.191066	1.82	0.069	.853898	64.53836
BlackMale		1.439738	.2779177	1.89	0.059	.9862186	2.101812
LatinoMale		.9310831	.160557	-0.41	0.679	.6640584	1.305481
OtherMale		1.916825	1.253688	0.99	0.320	.5319357	6.907258
-----							
3							
tenure		1.007322	.0049772	1.48	0.140	.9976137	1.017124
exper		1.073525	.008282	9.20	0.000	1.057415	1.089881
charter		1.11e-06	.0039834	-0.00	0.997	0	.
pyperstudlep		1.001908	.0048816	0.39	0.696	.9923859	1.011522
pyperstude~s		1.012243	.0032614	3.83	0.000	1.006058	1.018843
pyperstudw~e		1.026721	.0121732	2.22	0.026	1.003137	1.050859
pyperstudb~k		1.007395	.0130123	0.57	0.568	.9822117	1.033224
pyperstudl~o		1.007885	.0126068	0.63	0.530	.9834763	1.032899
pyperstudm~e		1.002328	.0043095	0.54	0.589	.9939168	1.01081
male		1.937921	.173592	7.39	0.000	1.62588	2.309849
age		.9531985	.0067623	-6.76	0.000	.9400364	.9665448
Asian		8.40e-07	.0007356	-0.02	0.987	0	.
Black		1.046602	.1731622	0.28	0.783	.7567437	1.447486
Latino		1.777542	.2832827	3.61	0.000	1.300666	2.429259
Other		.2195509	.2275669	-1.46	0.144	.0287909	1.674229
AsianMale		4811017	4.22e+09	0.02	0.986	0	.
BlackMale		1.431308	.329392	1.56	0.119	.9116812	2.247104
LatinoMale		.7874837	.1578503	-1.19	0.233	.5316419	1.166444
OtherMale		5.425053	6.145213	1.49	0.135	.5891257	49.95742
-----							

-----> pyschooltype = S

```

Iteration 0: log likelihood = -6808.9816
Iteration 1: log likelihood = -6499.8194
Iteration 2: log likelihood = -6488.0995
Iteration 3: log likelihood = -6487.8759
Iteration 4: log likelihood = -6487.8282
Iteration 5: log likelihood = -6487.8171
Iteration 6: log likelihood = -6487.8146
Iteration 7: log likelihood = -6487.814
Iteration 8: log likelihood = -6487.8139
Iteration 9: log likelihood = -6487.8139

```

Multinomial logistic regression	Number of obs	=	6954
	LR chi2(38)	=	642.34
	Prob > chi2	=	0.0000
Log likelihood = -6487.8139	Pseudo R2	=	0.0472

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					

tenure		1.003951	.0038992	1.02	0.310	.996338	1.011623
exper		1.002029	.0053775	0.38	0.706	.9915445	1.012624
charter		.3321966	.2725753	-1.34	0.179	.0665218	1.658924
pyperstudlep		1.021465	.004327	5.01	0.000	1.013019	1.029981
pyperstude~s		.9906754	.0022618	-4.10	0.000	.9862523	.9951185
pyperstudw~e		.9770149	.0067675	-3.36	0.001	.9638405	.9903694
pyperstudb~k		.98273	.0073031	-2.34	0.019	.96852	.9971485
pyperstudl~o		.9805496	.0071268	-2.70	0.007	.9666802	.9946179
pyperstudm~e		1.000362	.0023777	0.15	0.879	.9957123	1.005033
male		1.22496	.0845284	2.94	0.003	1.070002	1.402358
age		.9727633	.0047535	-5.65	0.000	.963491	.9821248
Asian		.2358706	.2484248	-1.37	0.170	.0299342	1.858577
Black		1.366743	.1716778	2.49	0.013	1.068481	1.748264
Latino		1.455034	.1815263	3.01	0.003	1.139409	1.858091
Other		.9094922	.4508387	-0.19	0.848	.3442351	2.402939
AsianMale		7.757967	9.394839	1.69	0.091	.7226856	83.28111
BlackMale		1.895473	.3234194	3.75	0.000	1.356686	2.648231
LatinoMale		1.28785	.1987027	1.64	0.101	.9517726	1.742599
OtherMale		1.518432	.9526422	0.67	0.506	.4439802	5.193102
-----							
3							
tenure		.992581	.0044492	-1.66	0.097	.9838989	1.00134
exper		1.050212	.0073875	6.96	0.000	1.035832	1.064791
charter		.8524306	.5122505	-0.27	0.790	.2625108	2.768031
pyperstudlep		1.009893	.0060406	1.65	0.100	.9981225	1.021802
pyperstude~s		1.010483	.0027005	3.90	0.000	1.005204	1.01579
pyperstudw~e		1.046059	.0130195	3.62	0.000	1.02085	1.071891
pyperstudb~k		1.031236	.0134975	2.35	0.019	1.005118	1.058033
pyperstudl~o		1.028568	.0131688	2.20	0.028	1.003078	1.054705
pyperstudm~e		1.017641	.0023244	7.66	0.000	1.013095	1.022207
male		1.9832	.1781897	7.62	0.000	1.662978	2.365085
age		.9715282	.0065821	-4.26	0.000	.9587128	.9845149
Asian		1.422078	1.115167	0.45	0.653	.3057926	6.613322
Black		.8006067	.1766	-1.01	0.313	.5195877	1.233615
Latino		1.425036	.2578362	1.96	0.050	.9995746	2.031591
Other		.843421	.6467378	-0.22	0.824	.1876465	3.790953
AsianMale		7.56e-07	.0005685	-0.02	0.985	0	.
BlackMale		2.227513	.5895983	3.03	0.002	1.325917	3.742174
LatinoMale		.9758067	.2070924	-0.12	0.908	.6437481	1.479148
OtherMale		2.295273	2.01561	0.95	0.344	.4105311	12.83284
-----							

### 2002-03

```
. mlogit admin tenure exper pyperstudlep pyperstudecodis pyperstudwhite pyperstudblack
pyperstudlatino pyperstudmobile male age Asian Black Latino O
> ther AsianMale BlackMale LatinoMale OtherMale if validcert2003a==1, rrr
```

```
Iteration 0: log likelihood = -24560.309
Iteration 1: log likelihood = -23544.944
Iteration 2: log likelihood = -23531.479
Iteration 3: log likelihood = -23531.469
Iteration 4: log likelihood = -23531.469
```

```
Multinomial logistic regression      Number of obs   =      23106
LR chi2(36)                        =      2057.68
Prob > chi2                        =      0.0000
Pseudo R2                          =      0.0419
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.008692	.0024227	3.60	0.000	1.003955 1.013451
exper		1.000306	.0031543	0.10	0.923	.994143 1.006508

pyperstudlep		1.009471	.0013763	6.91	0.000	1.006777	1.012172
pyperstude~s		.9915899	.0012128	-6.91	0.000	.9892157	.9939697
pyperstudw~e		.9881112	.0037868	-3.12	0.002	.9807171	.9955612
pyperstudb~k		.994245	.0041589	-1.38	0.168	.986127	1.00243
pyperstudl~o		.9928997	.003973	-1.78	0.075	.9851433	1.000717
pyperstudm~e		.9975862	.0013882	-1.74	0.082	.9948691	1.000311
male		1.216088	.0499931	4.76	0.000	1.121946	1.318128
age		.9694569	.0026185	-11.48	0.000	.9643383	.9746027
Asian		.5588385	.1897328	-1.71	0.087	.2872716	1.087126
Black		1.236664	.0800261	3.28	0.001	1.089355	1.403893
Latino		1.242796	.0744444	3.63	0.000	1.105127	1.397614
Other		.7434571	.1718305	-1.28	0.200	.4726321	1.169469
AsianMale		2.659466	1.317959	1.97	0.048	1.006844	7.024682
BlackMale		1.536886	.1553285	4.25	0.000	1.260704	1.873571
LatinoMale		1.318189	.1106217	3.29	0.001	1.118267	1.553853
OtherMale		1.536992	.5217277	1.27	0.205	.790194	2.989576

3							
tenure		1.013875	.0022528	6.20	0.000	1.009469	1.0183
exper		1.050538	.0035036	14.78	0.000	1.043694	1.057428
pyperstudlep		1.012877	.0014148	9.16	0.000	1.010108	1.015654
pyperstude~s		1.017948	.0012613	14.36	0.000	1.015479	1.020423
pyperstudw~e		1.012814	.0044913	2.87	0.004	1.004049	1.021655
pyperstudb~k		.9939462	.0047989	-1.26	0.209	.9845849	1.003397
pyperstudl~o		.9903931	.004572	-2.09	0.037	.9814725	.9993947
pyperstudm~e		1.003404	.0013489	2.53	0.011	1.000763	1.006051
male		1.311817	.0533533	6.67	0.000	1.211306	1.420668
age		.9741978	.0029399	-8.66	0.000	.9684526	.979977
Asian		.681595	.2497663	-1.05	0.296	.33236	1.397797
Black		.8319797	.061345	-2.49	0.013	.7200298	.9613356
Latino		1.196388	.0753073	2.85	0.004	1.05753	1.353478
Other		.5392751	.1463241	-2.28	0.023	.3168471	.9178485
AsianMale		1.690708	.925688	0.96	0.337	.5781263	4.944407
BlackMale		1.567589	.1789938	3.94	0.000	1.253252	1.960766
LatinoMale		.9559316	.0863965	-0.50	0.618	.8007478	1.14119
OtherMale		1.91415	.7280625	1.71	0.088	.9082694	4.034014

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=23106)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		40.838	2	0.000
exper		252.734	2	0.000
pyperstudlep		97.842	2	0.000
pyperstude~s		369.832	2	0.000
pyperstudw~e		28.249	2	0.000
pyperstudb~k		2.587	2	0.274
pyperstudl~o		5.615	2	0.060
pyperstudm~e		13.751	2	0.001
male		51.584	2	0.000
age		165.063	2	0.000
Asian		3.542	2	0.170
Black		25.063	2	0.000
Latino		15.959	2	0.000
Other		6.164	2	0.046
AsianMale		4.108	2	0.128
BlackMale		24.368	2	0.000
LatinoMale		14.637	2	0.001
OtherMale		3.592	2	0.166

\*\*\*\* Wald tests for independent variables (N=23106)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	40.720	2	0.000
exper	239.384	2	0.000
pyperstudlep	96.306	2	0.000
pyperstude~s	359.510	2	0.000
pyperstudw~e	27.948	2	0.000
pyperstudb~k	2.581	2	0.275
pyperstudl~o	5.607	2	0.061
pyperstudm~e	13.832	2	0.001
male	51.478	2	0.000
age	162.231	2	0.000
Asian	3.396	2	0.183
Black	25.012	2	0.000
Latino	15.937	2	0.000
Other	5.797	2	0.055
AsianMale	4.003	2	0.135
BlackMale	24.029	2	0.000
LatinoMale	14.645	2	0.001
OtherMale	3.549	2	0.170

\*\*\*\* Hausman tests of IIA assumption (N=23106)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-62.536	18	---	---
3	-19.290	18	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=23106)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-5223.460	-5211.742	23.436	19	0.219	for Ho
3	-5482.951	-5475.074	15.755	19	0.674	for Ho

\*\*\*\* Wald tests for combining alternatives (N=23106)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2-	3   1352.429	18	0.000
2-	1   608.254	18	0.000
3-	1   1007.611	18	0.000

\*\*\*\* LR tests for combining alternatives (N=23106)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2-	3   1493.488	18	0.000



```

2-      1 | 632.497  18    0.000
3-      1 | 1090.670  18    0.000
-----
by region:
. bysort region: mlogit admin tenure exper pyperstudlep pyperstuddecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile male age Asian
> Black Latino Other AsianMale BlackMale LatinoMale OtherMale if validcert2003a==1, rrr

-----
-----> region = 1

note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -1718.8996
Iteration 1: log likelihood = -1628.0134
Iteration 2: log likelihood = -1625.8564
Iteration 3: log likelihood = -1625.6309
Iteration 4: log likelihood = -1625.5955
Iteration 5: log likelihood = -1625.5873
Iteration 6: log likelihood = -1625.5853
Iteration 7: log likelihood = -1625.5849
Iteration 8: log likelihood = -1625.5848
Iteration 9: log likelihood = -1625.5848

Multinomial logistic regression                                Number of obs   =       1590
                                                                LR chi2(34)      =       186.63
                                                                Prob > chi2      =       0.0000
Log likelihood = -1625.5848                                    Pseudo R2       =       0.0543

-----
admin |          RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1      | (base outcome)
-----+-----
2      |
    tenure | 1.016374   .0092499     1.78   0.074   .9984052   1.034666
    exper  | 1.011984   .0135318     0.89   0.373   .9858064   1.038856
pyperstudlep | 1.007248   .0033165     2.19   0.028   1.000769   1.013769
pyperstud~s | 1.015603   .0068321     2.30   0.021   1.0023    1.029082
pyperstudw~e | .964602   .0682431    -0.51   0.610   .8397073   1.108073
pyperstudb~k | .8540418   .1030419    -1.31   0.191   .6741863   1.081878
pyperstudl~o | .9388771   .0623177    -0.95   0.342   .8243478   1.069318
pyperstudm~e | .9947113   .0056668    -0.93   0.352   .9836664   1.00588
    male   | 1.220884   .3367594     0.72   0.469   .7110259   2.096348
    age    | .9560144   .0109234    -3.94   0.000   .9348429   .9776654
    Asian  | 1.53e-07   .0002914    -0.01   0.993   0          .
    Black  | 9.34e-07   .0010271    -0.01   0.990   0          .
    Latino | .7385332   .1515242    -1.48   0.140   .4940031   1.104105
    Other  | .3713015   .4351981    -0.85   0.398   .0373288   3.69326
    AsianMale | 1.78e+13   4.93e+16     0.01   0.991   0          .
    BlackMale | .3629168   580.62     -0.00   0.999   0          .
    LatinoMale | 1.354478   .4175807     0.98   0.325   .7402008   2.478531
    OtherMale | (omitted)
-----
3      |
    tenure | 1.030013   .0095567     3.19   0.001   1.011451   1.048915
    exper  | 1.074257   .017495     4.40   0.000   1.040508   1.109099
pyperstudlep | 1.020101   .0036253     5.60   0.000   1.01302    1.027231
pyperstud~s | 1.003264   .0074664     0.44   0.661   .9887368   1.018005
pyperstudw~e | 1.017542   .083813     0.21   0.833   .8658457   1.195815
pyperstudb~k | 1.030146   .1042474     0.29   0.769   .8448114   1.256138
pyperstudl~o | .9999664   .077158     -0.00   1.000   .8596194   1.163227
pyperstudm~e | 1.006666   .005358     1.25   0.212   .9962189   1.017222
    male   | .742661   .2530413    -0.87   0.383   .3808615   1.448152
    age    | .9473974   .0139532    -3.67   0.000   .9204406   .9751438
    Asian  | 3.54e-07   .0007781    -0.01   0.995   0          .
    Black  | 3.44254    4.941282     0.86   0.389   .2065851   57.36659
    Latino | .8384463   .1867635    -0.79   0.429   .5418392   1.297418

```

Other		.5900503	.7103065	-0.44	0.661	.0557451	6.245556
AsianMale		8303650	3.24e+10	0.00	0.997	0	.
BlackMale		2.47e-07	.0002464	-0.02	0.988	0	.
LatinoMale		1.81997	.6801564	1.60	0.109	.8748922	3.785942
OtherMale		(omitted)					

-----  
-----> region = 2

Iteration 0: log likelihood = -800.63106  
Iteration 1: log likelihood = -760.02092  
Iteration 2: log likelihood = -758.34512  
Iteration 3: log likelihood = -758.05489  
Iteration 4: log likelihood = -757.98907  
Iteration 5: log likelihood = -757.97328  
Iteration 6: log likelihood = -757.97014  
Iteration 7: log likelihood = -757.96963  
Iteration 8: log likelihood = -757.96951  
Iteration 9: log likelihood = -757.96948

Multinomial logistic regression	Number of obs	=	814
	LR chi2(36)	=	85.32
	Prob > chi2	=	0.0000
Log likelihood = -757.96948	Pseudo R2	=	0.0533

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.01331	.0152991	0.88	0.381	.9837639 1.043744
exper		.9735664	.0194532	-1.34	0.180	.9361757 1.01245
pyperstudlep		.9920679	.0175162	-0.45	0.652	.9583239 1.027
pyperstude~s		1.004794	.0070677	0.68	0.497	.9910366 1.018742
pyperstudw~e		1.01352	.0381124	0.36	0.721	.9415074 1.09104
pyperstudb~k		1.034591	.048982	0.72	0.473	.9429074 1.135189
pyperstudl~o		1.008489	.0363373	0.23	0.815	.9397261 1.082284
pyperstudm~e		.9892515	.0084697	-1.26	0.207	.9727897 1.005992
male		1.184559	.3428308	0.59	0.558	.6717455 2.088855
age		.976906	.0158308	-1.44	0.149	.9463658 1.008432
Asian		3.59e+07	1.29e+11	0.00	0.996	0 .
Black		1.663593	1.679059	0.50	0.614	.230108 12.02714
Latino		1.050252	.2819683	0.18	0.855	.6205336 1.777548
Other		5.35e-07	.0004999	-0.02	0.988	0 .
AsianMale		1.89e-14	7.37e-11	-0.01	0.994	0 .
BlackMale		1.736566	2.207588	0.43	0.664	.1437514 20.97831
LatinoMale		1.899713	.7487545	1.63	0.104	.8773945 4.113214
OtherMale		1793809	1.68e+09	0.02	0.988	0 .
3						
tenure		.9968411	.0118416	-0.27	0.790	.9739001 1.020322
exper		1.039612	.0184867	2.18	0.029	1.004003 1.076484
pyperstudlep		.9865654	.014765	-0.90	0.366	.9580468 1.015933
pyperstude~s		1.033386	.0073365	4.63	0.000	1.019106 1.047866
pyperstudw~e		1.003679	.0320927	0.11	0.909	.942709 1.068592
pyperstudb~k		.9647662	.0425228	-0.81	0.416	.8849215 1.051815
pyperstudl~o		.9830022	.0299166	-0.56	0.573	.9260812 1.043422
pyperstudm~e		.9913965	.0085756	-1.00	0.318	.9747302 1.008348
male		1.279203	.3337797	0.94	0.345	.7670742 2.13325
age		.9773386	.0157144	-1.43	0.154	.9470192 1.008629
Asian		1.360309	8959.961	0.00	1.000	0 .
Black		3.370632	2.864529	1.43	0.153	.6372626 17.82806
Latino		.9541458	.2327468	-0.19	0.847	.5915336 1.539041
Other		8.72e-07	.00073	-0.02	0.987	0 .
AsianMale		8.71e-07	.0058566	-0.00	0.998	0 .

BlackMale		.5081421	.5849922	-0.59	0.556	.053217	4.851993
LatinoMale		1.244329	.4569424	0.60	0.552	.6058384	2.555721
OtherMale		1.805948	2018.407	0.00	1.000	0	.

-----  
-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -356.40628  
Iteration 1: log likelihood = -321.69075  
Iteration 2: log likelihood = -320.83474  
Iteration 3: log likelihood = -320.79017  
Iteration 4: log likelihood = -320.78039  
Iteration 5: log likelihood = -320.77783  
Iteration 6: log likelihood = -320.77795  
Iteration 7: log likelihood = -320.77791  
Iteration 8: log likelihood = -320.7779

Multinomial logistic regression	Number of obs	=	335
	LR chi2(30)	=	71.26
	Prob > chi2	=	0.0000
Log likelihood = -320.7779	Pseudo R2	=	0.1000

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.02189	.0224711	0.98	0.325	.9787834 1.066896
exper		.9381358	.0241596	-2.48	0.013	.891959 .9867031
pyperstudlep		.9735801	.0353911	-0.74	0.461	.9066283 1.045476
pyperstude~s		1.002111	.0141545	0.15	0.881	.9747487 1.03024
pyperstudw~e		.9819826	.0497053	-0.36	0.719	.8892386 1.0844
pyperstudb~k		.9920796	.0504692	-0.16	0.876	.8979333 1.096097
pyperstudl~o		.9950028	.0516044	-0.10	0.923	.8988308 1.101465
pyperstudm~e		.999818	.0121117	-0.02	0.988	.9763591 1.023841
male		1.003844	.3483185	0.01	0.991	.508525 1.981619
age		1.001338	.0238662	0.06	0.955	.9556368 1.049225
Asian		(omitted)				
Black		3.495986	2.394599	1.83	0.068	.9131316 13.38462
Latino		2.318142	1.302115	1.50	0.134	.7709293 6.970525
Other		1.89676	3052.004	0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		1.442027	2.067535	0.26	0.798	.0868053 23.95523
LatinoMale		.6977054	.5851725	-0.43	0.668	.134822 3.610633
OtherMale		(omitted)				
3						
tenure		1.019264	.0184372	1.05	0.291	.9837612 1.056049
exper		1.006553	.0263535	0.25	0.803	.9562042 1.059553
pyperstudlep		1.008247	.0314453	0.26	0.792	.9484612 1.071801
pyperstude~s		1.043018	.013456	3.26	0.001	1.016975 1.069728
pyperstudw~e		1.06588	.0561522	1.21	0.226	.9613146 1.181818
pyperstudb~k		1.061332	.0559803	1.13	0.259	.9570938 1.176924
pyperstudl~o		1.01946	.0548514	0.36	0.720	.9174274 1.13284
pyperstudm~e		1.014714	.0118447	1.25	0.211	.9917627 1.038197
male		1.406486	.4539358	1.06	0.291	.7471621 2.647623
age		1.009138	.0258703	0.35	0.723	.9596858 1.061138
Asian		(omitted)				
Black		2.080475	1.555357	0.98	0.327	.4806255 9.005714
Latino		1.297406	.8134097	0.42	0.678	.379677 4.433405
Other		2820124	2.67e+09	0.02	0.987	0 .
AsianMale		(omitted)				

BlackMale		.7222231	1.173987	-0.20	0.841	.0298557	17.4709
LatinoMale		1.490586	1.267395	0.47	0.639	.2815848	7.89051
OtherMale		(omitted)					

-----  
-----> region = 4

Iteration 0: log likelihood = -4957.5426  
Iteration 1: log likelihood = -4724.4629  
Iteration 2: log likelihood = -4718.8152  
Iteration 3: log likelihood = -4718.7988  
Iteration 4: log likelihood = -4718.7988

Multinomial logistic regression		Number of obs	=	4797
		LR chi2(36)	=	477.49
		Prob > chi2	=	0.0000
Log likelihood = -4718.7988		Pseudo R2	=	0.0482

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.012006	.0052311	2.31	0.021	1.001805 1.022311
exper		1.03212	.0072327	4.51	0.000	1.018041 1.046393
pyperstudlep		.9982264	.0033448	-0.53	0.596	.9916923 1.004804
pyperstude~s		.9932129	.0031954	-2.12	0.034	.9869698 .9994956
pyperstudw~e		1.001137	.0062772	0.18	0.856	.9889092 1.013516
pyperstudb~k		1.003618	.0074135	0.49	0.625	.9891925 1.018254
pyperstudl~o		1.008747	.0072302	1.22	0.224	.9946748 1.023018
pyperstudm~e		1.002387	.0033832	0.71	0.480	.9957778 1.00904
male		1.108104	.1010565	1.13	0.260	.9267289 1.324976
age		.9475585	.0056777	-8.99	0.000	.9364955 .9587522
Asian		.6325732	.3432276	-0.84	0.399	.2184025 1.832162
Black		.9116158	.0994605	-0.85	0.396	.73611 1.128966
Latino		1.299717	.1860647	1.83	0.067	.981731 1.720699
Other		.5889339	.2646127	-1.18	0.239	.2441265 1.420752
AsianMale		2.107752	1.83318	0.86	0.391	.3832678 11.59143
BlackMale		1.768447	.3021539	3.34	0.001	1.265195 2.471878
LatinoMale		1.195204	.2793061	0.76	0.445	.7560063 1.88955
OtherMale		.8737306	.812862	-0.15	0.885	.1410832 5.41103
3						
tenure		1.032052	.0058138	5.60	0.000	1.02072 1.04351
exper		1.065715	.0089826	7.55	0.000	1.048254 1.083467
pyperstudlep		1.013779	.0039424	3.52	0.000	1.006081 1.021535
pyperstude~s		1.018372	.0039941	4.64	0.000	1.010574 1.02623
pyperstudw~e		1.013908	.007937	1.76	0.078	.9984702 1.029584
pyperstudb~k		.9960688	.0093377	-0.42	0.674	.9779343 1.014539
pyperstudl~o		.9889355	.0088461	-1.24	0.214	.9717486 1.006426
pyperstudm~e		1.007297	.0037671	1.94	0.052	.9999402 1.014707
male		.9641851	.1033461	-0.34	0.734	.7814918 1.189588
age		.9622669	.0071807	-5.15	0.000	.9482953 .9764443
Asian		.7452754	.4527204	-0.48	0.628	.2265955 2.451221
Black		.5410127	.0734625	-4.52	0.000	.4145962 .7059754
Latino		1.195778	.1962345	1.09	0.276	.8668884 1.649445
Other		.5355614	.3023162	-1.11	0.269	.177141 1.619196
AsianMale		.6037941	.7728825	-0.39	0.693	.0491263 7.421019
BlackMale		2.073375	.4450302	3.40	0.001	1.361368 3.157767
LatinoMale		1.172679	.3356573	0.56	0.578	.6691738 2.055037
OtherMale		1.912363	1.916033	0.65	0.518	.268371 13.62716

-----  
-----> region = 5

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -555.1928  
 Iteration 1: log likelihood = -502.16288  
 Iteration 2: log likelihood = -500.32389  
 Iteration 3: log likelihood = -500.07437  
 Iteration 4: log likelihood = -500.04099  
 Iteration 5: log likelihood = -500.0369  
 Iteration 6: log likelihood = -500.03613  
 Iteration 7: log likelihood = -500.03596  
 Iteration 8: log likelihood = -500.03592  
 Iteration 9: log likelihood = -500.03591

Multinomial logistic regression	Number of obs	=	552
	LR chi2(34)	=	110.31
	Prob > chi2	=	0.0000
Log likelihood = -500.03591	Pseudo R2	=	0.0993

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9913602	.0152678	-0.56	0.573	.961883	1.021741
	exper	1.011213	.0235247	0.48	0.632	.9661407	1.058388
pyperstudlep		.9878875	.0390787	-0.31	0.758	.9141886	1.067528
pyperstude~s		1.013079	.0087884	1.50	0.134	.9959992	1.030451
pyperstudw~e		1.000541	.0417756	0.01	0.990	.9219227	1.085863
pyperstudb~k		.9874911	.0421202	-0.30	0.768	.9082935	1.073594
pyperstudl~o		1.005354	.0561113	0.10	0.924	.90118	1.121571
pyperstudm~e		.9943601	.014176	-0.40	0.692	.9669602	1.022536
	male	2.607997	.7290606	3.43	0.001	1.50784	4.510854
	age	.9620171	.0197501	-1.89	0.059	.924076	1.001516
	Asian	1.63e-07	.0004753	-0.01	0.996	0	.
	Black	2.832828	1.041271	2.83	0.005	1.378293	5.82236
	Latino	9667192	1.43e+10	0.01	0.991	0	.
	Other	3.29e-07	.000961	-0.01	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	.7938955	.4024914	-0.46	0.649	.293913	2.144411
	LatinoMale	1.34e-07	.0001986	-0.01	0.991	0	.
	OtherMale	7.65e+13	3.77e+17	0.01	0.995	0	.
3							
	tenure	.9963773	.0142255	-0.25	0.799	.9688824	1.024653
	exper	1.061643	.026223	2.42	0.015	1.011471	1.114303
pyperstudlep		.9933615	.0461423	-0.14	0.886	.9069189	1.088043
pyperstude~s		1.041973	.0092536	4.63	0.000	1.023994	1.060269
pyperstudw~e		1.00123	.0446716	0.03	0.978	.9173938	1.092727
pyperstudb~k		.9660081	.0439425	-0.76	0.447	.8836101	1.05609
pyperstudl~o		.9699712	.0604759	-0.49	0.625	.8583966	1.096048
pyperstudm~e		1.029758	.0106392	2.84	0.005	1.009115	1.050823
	male	3.807508	1.062602	4.79	0.000	2.203366	6.579529
	age	.9786958	.0222674	-0.95	0.344	.9360113	1.023327
	Asian	6.13e-07	.0018062	-0.00	0.996	0	.
	Black	1.922647	.8039084	1.56	0.118	.8472081	4.363238
	Latino	1.98e+07	2.92e+10	0.01	0.991	0	.
	Other	2.44e-06	.0071909	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	1.031256	.568037	0.06	0.955	.3503544	3.035469
	LatinoMale	2.37e-14	4.40e-11	-0.02	0.987	0	.
	OtherMale	361503.2	2.86e+09	0.00	0.999	0	.

-----  
 -----> region = 6

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -831.29754
Iteration 1: log likelihood = -780.60601
Iteration 2: log likelihood = -778.48327
Iteration 3: log likelihood = -778.32096
Iteration 4: log likelihood = -778.28731
Iteration 5: log likelihood = -778.28002
Iteration 6: log likelihood = -778.27881
Iteration 7: log likelihood = -778.27852
Iteration 8: log likelihood = -778.27845
Iteration 9: log likelihood = -778.27844
```

```
Multinomial logistic regression      Number of obs   =      768
                                     LR chi2(34)        =     106.04
                                     Prob > chi2         =     0.0000
Log likelihood = -778.27844          Pseudo R2        =     0.0638
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.974689	.0130779	-1.91	0.056	.9493909	1.000661
	exper	.9908518	.0168537	-0.54	0.589	.9583638	1.024441
pyperstudlep		1.012964	.0165034	0.79	0.429	.981129	1.045832
pyperstude~s		.9941751	.0072927	-0.80	0.426	.979984	1.008572
pyperstudw~e		.9891837	.0345937	-0.31	0.756	.9236528	1.059364
pyperstudb~k		.9882671	.0368041	-0.32	0.751	.9187021	1.0631
pyperstudl~o		.9923624	.0359245	-0.21	0.832	.9243916	1.065331
pyperstudm~e		.9748616	.0108816	-2.28	0.023	.9537656	.9964242
	male	1.135064	.220947	0.65	0.515	.7750484	1.662309
	age	.9901703	.0145918	-0.67	0.503	.9619799	1.019187
	Asian	1.28e-07	.0002928	-0.01	0.994	0	.
	Black	.9718945	.5328911	-0.05	0.959	.3318213	2.84665
	Latino	1.430564	.9241151	0.55	0.579	.4033185	5.074183
	Other	6.65e-07	.0012541	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	5.934788	5.210832	2.03	0.043	1.061788	33.17208
	LatinoMale	2.148099	1.924771	0.85	0.393	.3709782	12.43827
	OtherMale	2159515	4.07e+09	0.01	0.994	0	.
3							
	tenure	.9732998	.0119461	-2.20	0.027	.9501653	.9969975
	exper	1.071302	.0204582	3.61	0.000	1.031946	1.112159
pyperstudlep		1.014956	.0177297	0.85	0.395	.9807947	1.050308
pyperstude~s		1.019526	.0071092	2.77	0.006	1.005687	1.033556
pyperstudw~e		.996829	.03542	-0.09	0.929	.9297693	1.068725
pyperstudb~k		.9918417	.0367344	-0.22	0.825	.9223946	1.066517
pyperstudl~o		.9762193	.0352557	-0.67	0.505	.9095083	1.047823
pyperstudm~e		1.006152	.0065579	0.94	0.347	.9933806	1.019088
	male	1.879505	.3725665	3.18	0.001	1.274422	2.771875
	age	.9666802	.0167954	-1.95	0.051	.9343161	1.000165
	Asian	2.34e-07	.0005949	-0.01	0.995	0	.
	Black	1.336672	.6906346	0.56	0.574	.485539	3.679811
	Latino	.980728	.8604047	-0.02	0.982	.175703	5.474168
	Other	2.728547	3.89964	0.70	0.482	.1657272	44.92301
	AsianMale	(omitted)					
	BlackMale	1.700844	1.486901	0.61	0.543	.3065718	9.436191
	LatinoMale	.3219036	.4591974	-0.79	0.427	.0196554	5.271935
	OtherMale	1.35e-07	.0002482	-0.01	0.993	0	.

```
-----> region = 7
```

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -1014.8271
Iteration 1: log likelihood = -951.33208
Iteration 2: log likelihood = -948.51334
Iteration 3: log likelihood = -948.32092
Iteration 4: log likelihood = -948.29089
Iteration 5: log likelihood = -948.28371
Iteration 6: log likelihood = -948.28212
Iteration 7: log likelihood = -948.28177
Iteration 8: log likelihood = -948.28171

```

```

Multinomial logistic regression
Number of obs   =      1006
LR chi2(34)     =      133.09
Prob > chi2     =      0.0000
Pseudo R2      =      0.0656

Log likelihood = -948.28171

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.000381	.0133489	0.03	0.977	.974557	1.02689
	exper	.9881583	.0170383	-0.69	0.490	.9553218	1.022123
pypersistudlep		.9618835	.0234423	-1.59	0.111	.9170176	1.008945
pypersistude~s		.9884709	.0077657	-1.48	0.140	.973367	1.003809
pypersistudw~e		.9638203	.0834343	-0.43	0.670	.8134123	1.14204
pypersistudb~k		.9673732	.0848336	-0.38	0.705	.8146069	1.148788
pypersistudl~o		1.008085	.0899533	0.09	0.928	.8463362	1.200747
pypersistudm~e		.9949351	.0081157	-0.62	0.534	.979155	1.010969
	male	1.211596	.2489526	0.93	0.350	.8099487	1.812419
	age	.9759048	.0145008	-1.64	0.101	.9478936	1.004744
	Asian	3.76e-06	.0031235	-0.02	0.988	0	.
	Black	1.16963	.405792	0.45	0.652	.5925602	2.308686
	Latino	.3622252	.3990282	-0.92	0.357	.0418112	3.138083
	Other	3.44e-06	.0022749	-0.02	0.985	0	.
	AsianMale	(omitted)					
	BlackMale	4.99452	2.602594	3.09	0.002	1.798628	13.86904
	LatinoMale	7.381173	10.75925	1.37	0.170	.4239964	128.4957
	OtherMale	1.134885	952.8971	0.00	1.000	0	.
3							
	tenure	.9998688	.0099523	-0.01	0.989	.9805516	1.019567
	exper	1.018532	.0141105	1.33	0.185	.9912482	1.046567
pypersistudlep		1.050811	.0205352	2.54	0.011	1.011324	1.091841
pypersistude~s		1.023279	.0063337	3.72	0.000	1.01094	1.035768
pypersistudw~e		1.133465	.1123424	1.26	0.206	.9333451	1.376494
pypersistudb~k		1.113515	.1113471	1.08	0.282	.9153336	1.354606
pypersistudl~o		1.089378	.1102339	0.85	0.398	.8933996	1.328347
pypersistudm~e		1.000913	.0062506	0.15	0.884	.9887366	1.013239
	male	1.605681	.2678505	2.84	0.005	1.157889	2.22665
	age	.9908649	.0123868	-0.73	0.463	.9668823	1.015442
	Asian	1.774737	2.744757	0.37	0.711	.0856428	36.7771
	Black	1.006271	.312239	0.02	0.984	.5477629	1.848574
	Latino	.16472	.1787145	-1.66	0.096	.0196439	1.381227
	Other	1.852461	2.31617	0.49	0.622	.1597624	21.47948
	AsianMale	(omitted)					
	BlackMale	2.728908	1.353364	2.02	0.043	1.032398	7.213241
	LatinoMale	18.07595	25.91272	2.02	0.043	1.088589	300.1499
	OtherMale	1.248393	1.966916	0.14	0.888	.0569158	27.38227

```

-----
-----> region = 8

```

```

note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -352.29521
Iteration 1: log likelihood = -314.39725

```

```

Iteration 2: log likelihood = -312.54608
Iteration 3: log likelihood = -312.41324
Iteration 4: log likelihood = -312.3893
Iteration 5: log likelihood = -312.38409
Iteration 6: log likelihood = -312.38287
Iteration 7: log likelihood = -312.38259
Iteration 8: log likelihood = -312.38253
Iteration 9: log likelihood = -312.38252

```

```

Multinomial logistic regression
Log likelihood = -312.38252
Number of obs   =      348
LR chi2(34)     =      79.83
Prob > chi2     =      0.0000
Pseudo R2      =      0.1133

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.01163	.0251353	0.47	0.642	.9635464	1.062114
exper	.9655048	.028539	-1.19	0.235	.9111589	1.023092
pyperstudlep	1.014917	.0501776	0.30	0.765	.9211853	1.118186
pyperstude~s	.9871866	.0137511	-0.93	0.355	.9605996	1.014509
pyperstudw~e	.6520553	.114898	-2.43	0.015	.4616311	.92103
pyperstudb~k	.6829275	.1204067	-2.16	0.031	.4833921	.9648273
pyperstudl~o	.6377573	.1155415	-2.48	0.013	.447141	.9096333
pyperstudm~e	.9839949	.0156326	-1.02	0.310	.9538276	1.015116
male	2.613925	1.017162	2.47	0.014	1.219167	5.604319
age	1.020705	.029717	0.70	0.481	.9640913	1.080643
Asian	9.62e+07	7.33e+11	0.00	0.998	0	.
Black	4.10914	2.709664	2.14	0.032	1.128362	14.96419
Latino	12.24745	18.67231	1.64	0.100	.6170409	243.0956
Other	3.91e-07	.0006945	-0.01	0.993	0	.
AsianMale	(omitted)					
BlackMale	.2364517	.2682513	-1.27	0.204	.0255897	2.184838
LatinoMale	.1519481	689.5212	-0.00	1.000	0	.
OtherMale	4122104	7.33e+09	0.01	0.993	0	.
3						
tenure	1.044885	.0192143	2.39	0.017	1.007896	1.083231
exper	.9749516	.0218039	-1.13	0.257	.9331398	1.018637
pyperstudlep	.9917248	.0307827	-0.27	0.789	.9331904	1.053931
pyperstude~s	1.03222	.0114153	2.87	0.004	1.010087	1.054838
pyperstudw~e	.9694309	.1509328	-0.20	0.842	.7144822	1.315353
pyperstudb~k	.9428757	.1465001	-0.38	0.705	.6953417	1.278529
pyperstudl~o	.9427496	.1477878	-0.38	0.707	.6933616	1.281837
pyperstudm~e	.9915341	.0134494	-0.63	0.531	.9655212	1.018248
male	1.692117	.4716179	1.89	0.059	.9799144	2.92195
age	1.046906	.0233927	2.05	0.040	1.002046	1.093773
Asian	.3514967	4130.196	-0.00	1.000	0	.
Black	1.644943	.9723093	0.84	0.400	.5164329	5.239472
Latino	4.887475	5.963264	1.30	0.193	.4472225	53.41281
Other	4.46e-07	.0005453	-0.01	0.990	0	.
AsianMale	(omitted)					
BlackMale	.9102056	.8684932	-0.10	0.921	.1402641	5.906534
LatinoMale	2222852	5.03e+09	0.01	0.995	0	.
OtherMale	3127720	3.82e+09	0.01	0.990	0	.

```

-----> region = 9

```

```

note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -225.69419
Iteration 1: log likelihood = -204.23621

```



```

Iteration 2: log likelihood = -202.44018
Iteration 3: log likelihood = -202.05129
Iteration 4: log likelihood = -201.9672
Iteration 5: log likelihood = -201.94849
Iteration 6: log likelihood = -201.94478
Iteration 7: log likelihood = -201.94394
Iteration 8: log likelihood = -201.94373
Iteration 9: log likelihood = -201.94369
Iteration 10: log likelihood = -201.94368

```

Multinomial logistic regression

```

Number of obs   =      213
LR chi2(32)     =      47.50
Prob > chi2     =      0.0382
Pseudo R2      =      0.1052

```

Log likelihood = -201.94368

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1						
tenure	.9717299	.0229988	-1.21	0.226	.9276827	1.017869
exper	1.008463	.0380868	0.22	0.823	.9365104	1.085944
pypersistudlep	.9959334	.0631545	-0.06	0.949	.8795359	1.127735
pypersistude~s	.9711087	.0147369	-1.93	0.053	.9426502	1.000426
pypersistudw~e	.9784695	.1157407	-0.18	0.854	.7759983	1.233769
pypersistudb~k	.994401	.1260449	-0.04	0.965	.7756537	1.274839
pypersistudl~o	.9979583	.1208155	-0.02	0.987	.7871611	1.265206
pypersistudm~e	.9990855	.0152689	-0.06	0.952	.9696028	1.029465
male	.8229176	.2871676	-0.56	0.576	.4152587	1.630775
age	1.025788	.0311353	0.84	0.402	.9665434	1.088663
Asian	6.23e-08	.0001818	-0.01	0.995	0	.
Black	.4247097	.536912	-0.68	0.498	.0356464	5.060217
Latino	4483295	8.51e+09	0.01	0.994	0	.
Other	2.77e-07	.0003654	-0.01	0.991	0	.
AsianMale	(omitted)					
BlackMale	7603147	9.05e+09	0.01	0.989	0	.
LatinoMale	8.66e-14	1.99e-10	-0.01	0.990	0	.
OtherMale	(omitted)					
2						
tenure	.9961589	.0343447	-0.11	0.911	.9310686	1.0658
exper	.895737	.0422603	-2.33	0.020	.8166225	.9825161
pypersistudlep	.9901205	.0714347	-0.14	0.891	.8595596	1.140513
pypersistude~s	.9780174	.0190475	-1.14	0.254	.9413886	1.016071
pypersistudw~e	.7772519	.1074208	-1.82	0.068	.5928171	1.019067
pypersistudb~k	.7983623	.1170228	-1.54	0.124	.5990073	1.064064
pypersistudl~o	.8037437	.1138449	-1.54	0.123	.6089064	1.060925
pypersistudm~e	1.012307	.0174668	0.71	0.478	.9786452	1.047127
male	.588573	.259844	-1.20	0.230	.2477483	1.398266
age	1.061518	.0356379	1.78	0.075	.9939171	1.133716
Asian	1.03e-07	.000394	-0.00	0.997	0	.
Black	.471573	.5286212	-0.67	0.503	.0524052	4.243497
Latino	1.25e+07	2.38e+10	0.01	0.993	0	.
Other	.8505089	1.102537	-0.12	0.901	.0670261	10.79229
AsianMale	(omitted)					
BlackMale	1.02e+07	1.22e+10	0.01	0.989	0	.
LatinoMale	1.51e-07	.0002867	-0.01	0.993	0	.
OtherMale	(omitted)					
3	(base outcome)					

```

-----
-----> region = 10

```

```

Iteration 0: log likelihood = -3526.8049
Iteration 1: log likelihood = -3351.6194
Iteration 2: log likelihood = -3349.1013

```

Iteration 3: log likelihood = -3349.0975  
 Iteration 4: log likelihood = -3349.0975

Multinomial logistic regression	Number of obs	=	3271
	LR chi2(36)	=	355.41
	Prob > chi2	=	0.0000
Log likelihood = -3349.0975	Pseudo R2	=	0.0504

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.006857	.0063344	1.09	0.277	.9945184	1.019349
	exper	1.002455	.0081905	0.30	0.764	.9865301	1.018638
pyperstudlep		.9988869	.0050943	-0.22	0.827	.9889521	1.008922
pyperstude~s		.9955369	.0036592	-1.22	0.224	.9883907	1.002735
pyperstudw~e		.980175	.0082267	-2.39	0.017	.9641829	.9964324
pyperstudb~k		.9819607	.0088023	-2.03	0.042	.9648592	.9993653
pyperstudl~o		.9844863	.0085572	-1.80	0.072	.9678566	1.001402
pyperstudm~e		.996537	.0042576	-0.81	0.417	.9882271	1.004917
	male	1.30197	.1372901	2.50	0.012	1.058872	1.600879
	age	.9671734	.0067465	-4.78	0.000	.9540405	.9804871
	Asian	.9093158	.8500481	-0.10	0.919	.1455436	5.681151
	Black	1.423972	.2100929	2.40	0.017	1.066388	1.901463
	Latino	1.411381	.2819524	1.72	0.085	.9541116	2.087802
	Other	.6958158	.373108	-0.68	0.499	.2432573	1.990319
	AsianMale	1.59729	1.820509	0.41	0.681	.1710914	14.91213
	BlackMale	1.540097	.3508401	1.90	0.058	.9854649	2.406884
	LatinoMale	1.962339	.6349168	2.08	0.037	1.040799	3.699825
	OtherMale	2.059917	1.45799	1.02	0.307	.5144904	8.247497
3							
	tenure	1.020196	.0061709	3.31	0.001	1.008172	1.032362
	exper	1.057181	.0097953	6.00	0.000	1.038155	1.076554
pyperstudlep		1.017923	.0058134	3.11	0.002	1.006593	1.029381
pyperstude~s		1.015208	.0038226	4.01	0.000	1.007744	1.022728
pyperstudw~e		1.008614	.0092932	0.93	0.352	.9905632	1.026994
pyperstudb~k		.991605	.0098773	-0.85	0.397	.9724336	1.011154
pyperstudl~o		.9755922	.0092984	-2.59	0.010	.9575367	.9939881
pyperstudm~e		1.001758	.0044282	0.40	0.691	.9931162	1.010475
	male	.9978149	.1122726	-0.02	0.984	.8003391	1.244016
	age	.9692369	.0080788	-3.75	0.000	.9535314	.9852012
	Asian	2.22334	1.926737	0.92	0.357	.4067788	12.15216
	Black	1.001827	.1694498	0.01	0.991	.7191502	1.395615
	Latino	1.767399	.3633571	2.77	0.006	1.181238	2.644429
	Other	.3803157	.2505157	-1.47	0.142	.1045813	1.38304
	AsianMale	.9762164	1.059567	-0.02	0.982	.1163237	8.192643
	BlackMale	2.021177	.5198765	2.74	0.006	1.220852	3.346154
	LatinoMale	2.289564	.7879718	2.41	0.016	1.166286	4.494699
	OtherMale	2.106692	1.900901	0.83	0.409	.3593755	12.34962

-----> region = 11

Iteration 0: log likelihood = -2487.1757  
 Iteration 1: log likelihood = -2341.9189  
 Iteration 2: log likelihood = -2339.4291  
 Iteration 3: log likelihood = -2339.4258  
 Iteration 4: log likelihood = -2339.4258

Multinomial logistic regression	Number of obs	=	2294
	LR chi2(36)	=	295.50
	Prob > chi2	=	0.0000
Log likelihood = -2339.4258	Pseudo R2	=	0.0594

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.001457	.0078424	0.19	0.853	.9862031	1.016946
	exper	.9978954	.0097874	-0.21	0.830	.9788956	1.017264
pyperstudlep		1.000967	.007558	0.13	0.898	.9862622	1.01589
pyperstude~s		.9965234	.0038599	-0.90	0.369	.9889868	1.004117
pyperstudw~e		1.003473	.0128081	0.27	0.786	.9786813	1.028893
pyperstudb~k		1.014954	.0139669	1.08	0.281	.9879457	1.042701
pyperstudl~o		1.014641	.0132396	1.11	0.265	.9890213	1.040925
pyperstudm~e		.9877881	.0040914	-2.97	0.003	.9798016	.9958397
	male	1.049876	.1223664	0.42	0.676	.8354637	1.319314
	age	.9731366	.0081722	-3.24	0.001	.9572504	.9892863
	Asian	.3183065	.3573171	-1.02	0.308	.0352633	2.873212
	Black	1.481791	.3261904	1.79	0.074	.9625198	2.281204
	Latino	2.087883	.6776925	2.27	0.023	1.105148	3.944502
	Other	1.000618	.5818413	0.00	0.999	.3201178	3.127711
	AsianMale	5.029374	7.328129	1.11	0.268	.2892393	87.45217
	BlackMale	1.261733	.4316141	0.68	0.497	.6453389	2.466873
	LatinoMale	.8244371	.3698782	-0.43	0.667	.3421934	1.986294
	OtherMale	.5313653	.4800079	-0.70	0.484	.0904611	3.121221
3							
	tenure	1.004461	.0071329	0.63	0.531	.9905771	1.018539
	exper	1.064253	.0110718	5.99	0.000	1.042772	1.086176
pyperstudlep		1.013109	.0082651	1.60	0.110	.997039	1.029439
pyperstude~s		1.022986	.0040592	5.73	0.000	1.015061	1.030973
pyperstudw~e		1.028062	.0143294	1.99	0.047	1.000357	1.056534
pyperstudb~k		1.015209	.0153176	1.00	0.317	.9856266	1.045679
pyperstudl~o		1.000365	.0144645	0.03	0.980	.9724132	1.02912
pyperstudm~e		.9951228	.0038046	-1.28	0.201	.9876938	1.002608
	male	1.277765	.1526688	2.05	0.040	1.010994	1.61493
	age	.9833171	.0093115	-1.78	0.076	.9652353	1.001738
	Asian	1.068663	.9507186	0.07	0.940	.1868888	6.110804
	Black	1.297586	.3366009	1.00	0.315	.7804227	2.157459
	Latino	3.457764	1.132279	3.79	0.000	1.819956	6.569462
	Other	.5993711	.415366	-0.74	0.460	.1541028	2.331209
	AsianMale	1.054645	1.613807	0.03	0.972	.052554	21.16442
	BlackMale	1.028696	.4181654	0.07	0.945	.4637389	2.281923
	LatinoMale	.4503525	.2190217	-1.64	0.101	.1736138	1.16821
	OtherMale	1.590135	1.51535	0.49	0.626	.2456209	10.29444

-----  
-----> region = 12

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -889.00032  
Iteration 1: log likelihood = -833.64968  
Iteration 2: log likelihood = -831.67391  
Iteration 3: log likelihood = -831.49892  
Iteration 4: log likelihood = -831.46799  
Iteration 5: log likelihood = -831.46122  
Iteration 6: log likelihood = -831.4596  
Iteration 7: log likelihood = -831.45925  
Iteration 8: log likelihood = -831.45918  
Iteration 9: log likelihood = -831.45916

Multinomial logistic regression	Number of obs	=	830
	LR chi2(32)	=	115.08
	Prob > chi2	=	0.0000
Log likelihood = -831.45916	Pseudo R2	=	0.0647

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9975952	.0126636	-0.19	0.850	.9730812	1.022727
	exper	.9964822	.0165735	-0.21	0.832	.9645225	1.029501
pyperstudlep		1.01007	.0244144	0.41	0.678	.9633341	1.059072
pyperstude~s		.9869882	.0075482	-1.71	0.087	.9723043	1.001894
pyperstudw~e		.8940907	.043196	-2.32	0.020	.8133129	.9828912
pyperstudb~k		.9251004	.047763	-1.51	0.132	.8360674	1.023615
pyperstudl~o		.9140627	.0442155	-1.86	0.063	.8313831	1.004965
pyperstudm~e		.9914212	.0079555	-1.07	0.283	.9759507	1.007137
	male	.9663085	.203659	-0.16	0.871	.6393176	1.460545
	age	.9696748	.0144609	-2.06	0.039	.9417422	.9984358
	Asian	(omitted)					
	Black	1.160639	.4039802	0.43	0.669	.5867083	2.296003
	Latino	1.795677	1.237494	0.85	0.396	.4651802	6.93163
	Other	.1698365	.2048216	-1.47	0.142	.0159767	1.805403
	AsianMale	(omitted)					
	BlackMale	1.060442	.6584643	0.09	0.925	.3140126	3.581185
	LatinoMale	.5507203	.5041341	-0.65	0.515	.0915659	3.31229
	OtherMale	6.95e-06	.0057563	-0.01	0.989	0	.
3							
	tenure	1.004123	.0108244	0.38	0.703	.9831303	1.025565
	exper	1.026511	.0157423	1.71	0.088	.9961154	1.057833
pyperstudlep		1.081128	.0238767	3.53	0.000	1.035329	1.128953
pyperstude~s		.99983	.0065854	-0.03	0.979	.9870058	1.012821
pyperstudw~e		1.037053	.0528999	0.71	0.476	.9383855	1.146095
pyperstudb~k		1.041394	.0563983	0.75	0.454	.9365194	1.158012
pyperstudl~o		1.013925	.0517804	0.27	0.787	.9173507	1.120665
pyperstudm~e		1.007592	.0067838	1.12	0.261	.9943832	1.020976
	male	1.36563	.2521539	1.69	0.091	.9509654	1.961108
	age	.9852252	.0137759	-1.06	0.287	.9585915	1.012599
	Asian	(omitted)					
	Black	.9253847	.3562598	-0.20	0.840	.435133	1.967989
	Latino	.7140907	.5914615	-0.41	0.684	.1408403	3.620595
	Other	1.55e-06	.0008773	-0.02	0.981	0	.
	AsianMale	(omitted)					
	BlackMale	1.517517	.9278584	0.68	0.495	.4578062	5.030202
	LatinoMale	.2789198	.3825163	-0.93	0.352	.0189725	4.100477
	OtherMale	1170521	6.62e+08	0.02	0.980	0	.

-----  
-----> region = 13

Iteration 0: log likelihood = -1529.011  
Iteration 1: log likelihood = -1461.5737  
Iteration 2: log likelihood = -1460.0747  
Iteration 3: log likelihood = -1459.8424  
Iteration 4: log likelihood = -1459.8049  
Iteration 5: log likelihood = -1459.7962  
Iteration 6: log likelihood = -1459.7942  
Iteration 7: log likelihood = -1459.7938  
Iteration 8: log likelihood = -1459.7937  
Iteration 9: log likelihood = -1459.7937

Multinomial logistic regression

Number of obs = 1416  
LR chi2(36) = 138.43  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0453

Log likelihood = -1459.7937

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.022866	.0105318	2.20	0.028	1.002431	1.043717
	exper	.9649718	.011352	-3.03	0.002	.9429768	.9874798
pyperstudlep		.991304	.0088996	-0.97	0.331	.9740137	1.008901
pyperstude~s		.9922406	.0068462	-1.13	0.259	.9789126	1.00575
pyperstudw~e		.969606	.021069	-1.42	0.155	.9291785	1.011792
pyperstudb~k		.9813051	.0244376	-0.76	0.449	.9345584	1.03039
pyperstudl~o		.9884388	.0224153	-0.51	0.608	.9454677	1.033363
pyperstudm~e		.9849802	.0065403	-2.28	0.023	.9722444	.9978827
	male	1.073916	.1727757	0.44	0.658	.7834764	1.472024
	age	.9937311	.0102043	-0.61	0.540	.973931	1.013934
	Asian	.565698	.6742439	-0.48	0.633	.0547085	5.849441
	Black	2.099293	.6230872	2.50	0.012	1.173355	3.755922
	Latino	1.130171	.2641461	0.52	0.601	.7148247	1.786853
	Other	2.950178	3.660935	0.87	0.383	.2591645	33.58312
	AsianMale	1.946492	3.619127	0.36	0.720	.050887	74.45572
	BlackMale	1.080823	.5368609	0.16	0.876	.4082724	2.861272
	LatinoMale	2.050318	.7935904	1.86	0.064	.9601905	4.378092
	OtherMale	.4618793	.6645229	-0.54	0.591	.0275343	7.747894
3							
	tenure	1.020443	.0096571	2.14	0.032	1.00169	1.039548
	exper	1.00671	.0118218	0.57	0.569	.9838042	1.030149
pyperstudlep		1.000589	.0086359	0.07	0.946	.9838051	1.017659
pyperstude~s		1.028461	.006792	4.25	0.000	1.015235	1.04186
pyperstudw~e		.9537683	.0210283	-2.15	0.032	.9134313	.9958866
pyperstudb~k		.9254497	.0238402	-3.01	0.003	.8798837	.9733754
pyperstudl~o		.9376338	.0214727	-2.81	0.005	.8964785	.9806783
pyperstudm~e		.9939749	.0058531	-1.03	0.305	.9825689	1.005513
	male	1.062489	.1662135	0.39	0.698	.7819233	1.443725
	age	1.003701	.0108146	0.34	0.732	.982727	1.025123
	Asian	.8855153	1.035577	-0.10	0.917	.0894844	8.76284
	Black	.8207343	.30226	-0.54	0.592	.3987697	1.689208
	Latino	.9951354	.239373	-0.02	0.984	.621057	1.594531
	Other	4.270006	5.055452	1.23	0.220	.4194234	43.47147
	AsianMale	7.60e-07	.0014451	-0.01	0.994	0	.
	BlackMale	2.702901	1.506769	1.78	0.074	.9063925	8.060167
	LatinoMale	1.433641	.6015655	0.86	0.391	.6298982	3.262949
	OtherMale	1.42e-07	.0001342	-0.02	0.987	0	.

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -261.48627  
Iteration 1: log likelihood = -232.28337  
Iteration 2: log likelihood = -228.56359  
Iteration 3: log likelihood = -228.28928  
Iteration 4: log likelihood = -228.24515  
Iteration 5: log likelihood = -228.23797  
Iteration 6: log likelihood = -228.23627  
Iteration 7: log likelihood = -228.2359  
Iteration 8: log likelihood = -228.23581  
Iteration 9: log likelihood = -228.2358

Multinomial logistic regression

Number of obs = 278  
LR chi2(30) = 66.50  
Prob > chi2 = 0.0001  
Pseudo R2 = 0.1272

Log likelihood = -228.2358

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9875852	.0312597	-0.39	0.693	.9281792	1.050793
	exper	1.019347	.0428932	0.46	0.649	.9386513	1.10698
pyperstudlep		1.028513	.084173	0.34	0.731	.8760886	1.207457
pyperstude~s		.985528	.0193397	-0.74	0.458	.9483425	1.024172
pyperstudw~e		.7615601	.2024304	-1.02	0.305	.4523204	1.282219
pyperstudb~k		.7738612	.2261476	-0.88	0.380	.4364294	1.372183
pyperstudl~o		.7772752	.2056647	-0.95	0.341	.4627533	1.30557
pyperstudm~e		.9672464	.0380618	-0.85	0.397	.8954509	1.044798
	male	4.498105	2.510147	2.69	0.007	1.506678	13.42885
	age	1.007962	.0394495	0.20	0.839	.9335331	1.088324
	Asian	(omitted)					
	Black	7.89e-06	.012457	-0.01	0.994	0	.
	Latino	7.020514	9.399715	1.46	0.146	.5089807	96.83592
	Other	5.09e-07	.0015952	-0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	126114.1	4.08e+08	0.00	0.997	0	.
	LatinoMale	.4615015	.7028681	-0.51	0.612	.0233233	9.13178
	OtherMale	(omitted)					
3							
	tenure	1.02424	.0207388	1.18	0.237	.9843887	1.065705
	exper	1.066085	.0330166	2.07	0.039	1.003298	1.1328
pyperstudlep		1.008247	.0473293	0.17	0.861	.9196227	1.105412
pyperstude~s		1.029449	.0133147	2.24	0.025	1.003681	1.055879
pyperstudw~e		1.012618	.2065665	0.06	0.951	.6789015	1.510373
pyperstudb~k		.9112104	.2066761	-0.41	0.682	.5841891	1.421294
pyperstudl~o		1.015615	.2067714	0.08	0.939	.6814451	1.513656
pyperstudm~e		.9958526	.0176956	-0.23	0.815	.9617668	1.031146
	male	1.814387	.5544638	1.95	0.051	.9968019	3.302563
	age	.9548267	.0262162	-1.68	0.092	.9048019	1.007617
	Asian	(omitted)					
	Black	8.199656	10.72962	1.61	0.108	.6308992	106.5691
	Latino	2.207259	2.22603	0.79	0.432	.3057827	15.93286
	Other	2.01e-07	.0003365	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	654200.8	7.52e+08	0.01	0.991	0	.
	LatinoMale	.3521912	.4776981	-0.77	0.442	.0246742	5.027063
	OtherMale	(omitted)					

-----> region = 15

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -294.19477  
Iteration 1: log likelihood = -264.61151  
Iteration 2: log likelihood = -261.60194  
Iteration 3: log likelihood = -261.46775  
Iteration 4: log likelihood = -261.44025  
Iteration 5: log likelihood = -261.43398  
Iteration 6: log likelihood = -261.43243  
Iteration 7: log likelihood = -261.43213  
Iteration 8: log likelihood = -261.43208  
Iteration 9: log likelihood = -261.43207

Multinomial logistic regression	Number of obs	=	289
	LR chi2(28)	=	65.53
	Prob > chi2	=	0.0001

Pseudo R2 = 0.1114

[illegible]

Log likelihood = -452.07345

Pseudo R2 = 0.1059

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.071209	.0237824	3.10	0.002	1.025596	1.11885
exper	.9033134	.0254951	-3.60	0.000	.854701	.9546908
pyperstudlep	1.018087	.0193199	0.94	0.345	.9809163	1.056666
pyperstude~s	.9756227	.0087005	-2.77	0.006	.9587181	.9928254
pyperstudw~e	.9572241	.0315833	-1.32	0.185	.897281	1.021172
pyperstudb~k	1.003459	.0377806	0.09	0.927	.9320763	1.080308
pyperstudl~o	.9715308	.0311533	-0.90	0.368	.9123507	1.03455
pyperstudm~e	1.030165	.014685	2.08	0.037	1.001781	1.059353
male	1.171778	.3308582	0.56	0.575	.6737577	2.037919
age	1.011167	.0226316	0.50	0.620	.9677689	1.056512
Asian	(omitted)					
Black	.677422	.7011911	-0.38	0.707	.0890822	5.151426
Latino	3.475713	2.061189	2.10	0.036	1.087074	11.11294
Other	1.259786	1.338011	0.22	0.828	.1571235	10.10072
AsianMale	(omitted)					
BlackMale	3.99e-07	.0002865	-0.02	0.984	0	.
LatinoMale	.5128455	.4747044	-0.72	0.471	.0835777	3.1469
OtherMale	2589873	2.42e+09	0.02	0.987	0	.
3						
tenure	1.036412	.0162531	2.28	0.023	1.005041	1.068762
exper	1.037208	.0248519	1.52	0.127	.9896248	1.087078
pyperstudlep	1.046976	.0181733	2.64	0.008	1.011956	1.083208
pyperstude~s	1.008068	.0073835	1.10	0.273	.9937004	1.022644
pyperstudw~e	1.052188	.0403709	1.33	0.185	.9759639	1.134364
pyperstudb~k	1.041294	.0442176	0.95	0.341	.958138	1.131668
pyperstudl~o	1.035043	.0387752	0.92	0.358	.9617676	1.1139
pyperstudm~e	1.01657	.0144974	1.15	0.249	.9885488	1.045385
male	1.9121	.475277	2.61	0.009	1.174718	3.112345
age	.9769412	.021962	-1.04	0.299	.934831	1.020948
Asian	(omitted)					
Black	.476592	.5834814	-0.61	0.545	.0432553	5.251153
Latino	.8046079	.6358653	-0.28	0.783	.170961	3.786793
Other	.5147164	.615249	-0.56	0.578	.0494439	5.358254
AsianMale	(omitted)					
BlackMale	.9917383	1.780691	-0.00	0.996	.0293794	33.47735
LatinoMale	.4545784	.531986	-0.67	0.501	.0458628	4.505642
OtherMale	427042.1	4.00e+08	0.01	0.989	0	.

-----> region = 17

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -465.45976  
Iteration 1: log likelihood = -420.8015  
Iteration 2: log likelihood = -419.30721  
Iteration 3: log likelihood = -419.15785  
Iteration 4: log likelihood = -419.13102  
Iteration 5: log likelihood = -419.12579  
Iteration 6: log likelihood = -419.12466  
Iteration 7: log likelihood = -419.1244  
Iteration 8: log likelihood = -419.12434  
Iteration 9: log likelihood = -419.12432

Multinomial logistic regression

Number of obs = 458  
LR chi2(32) = 92.67  
Prob > chi2 = 0.0000



Log likelihood = -419.12432                      Pseudo R2                      =                      0.0995

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.0338	.0237811	1.45	0.148	.9882245	1.081476
	exper	.9993742	.0299924	-0.02	0.983	.9422856	1.059922
pyp	erstudlep	.9894275	.0399028	-0.26	0.792	.9142305	1.07081
pyp	erstud~s	.9741058	.0111687	-2.29	0.022	.9524596	.9962438
pyp	erstudw~e	1.093208	.1681729	0.58	0.562	.8086462	1.477906
pyp	erstudb~k	1.106901	.1698377	0.66	0.508	.8194156	1.495247
pyp	erstudl~o	1.117347	.1710114	0.72	0.468	.8277719	1.508223
pyp	erstudm~e	.9930112	.0107792	-0.65	0.518	.9721075	1.014364
	male	1.756175	.5507263	1.80	0.073	.9498086	3.24713
	age	.9758759	.0256926	-0.93	0.354	.9267966	1.027554
	Asian	3.81e-07	.0013325	-0.00	0.997	0	.
	Black	2.834898	2.859855	1.03	0.302	.3925027	20.4754
	Latino	.5066064	.550584	-0.63	0.532	.0601974	4.263473
	Other	1.02e+07	1.94e+10	0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	4169646	7.88e+09	0.01	0.994	0	.
	LatinoMale	9.806516	12.94197	1.73	0.084	.7381707	130.2785
	OtherMale	(omitted)					
3							
	tenure	1.007066	.0158694	0.45	0.655	.9764381	1.038655
	exper	1.072553	.025689	2.92	0.003	1.023367	1.124103
pyp	erstudlep	1.023682	.0256424	0.93	0.350	.9746376	1.075194
pyp	erstud~s	1.022213	.0091744	2.45	0.014	1.004388	1.040353
pyp	erstudw~e	1.21936	.1589895	1.52	0.128	.9443777	1.574411
pyp	erstudb~k	1.171541	.1536012	1.21	0.227	.9060583	1.514812
pyp	erstudl~o	1.200981	.15662	1.40	0.160	.930103	1.550749
pyp	erstudm~e	.9989556	.0089971	-0.12	0.908	.9814764	1.016746
	male	2.626215	.6469991	3.92	0.000	1.620416	4.256316
	age	.9493319	.0212408	-2.32	0.020	.9086004	.9918893
	Asian	5.30e-07	.0011817	-0.01	0.995	0	.
	Black	2.412786	2.397155	0.89	0.375	.3442167	16.91242
	Latino	1.65561	.930897	0.90	0.370	.5499897	4.983812
	Other	2746423	5.22e+09	0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	4618081	8.72e+09	0.01	0.994	0	.
	LatinoMale	1.845305	1.652089	0.68	0.494	.3191489	10.66948
	OtherMale	(omitted)					

```
-----> region = 18
```

```
note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0:    log likelihood = -418.50231
Iteration 1:    log likelihood = -381.64399
Iteration 2:    log likelihood = -380.30382
Iteration 3:    log likelihood = -380.2554
Iteration 4:    log likelihood = -380.24387
Iteration 5:    log likelihood = -380.24151
Iteration 6:    log likelihood = -380.24101
Iteration 7:    log likelihood = -380.2409
Iteration 8:    log likelihood = -380.24087
```

Multinomial logistic regression	Number of obs	=	399
	LR chi2(32)	=	76.52
	Prob > chi2	=	0.0000
Log likelihood = -380.24087	Pseudo R2	=	0.0914

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9963511	.02283	-0.16	0.873	.9525951	1.042117
	exper	.9665208	.0256833	-1.28	0.200	.9174709	1.018193
pyperstudlep		1.030483	.0191613	1.61	0.106	.9936033	1.068731
pyperstude~s		.9742643	.0143713	-1.77	0.077	.9465002	1.002843
pyperstudw~e		.7780815	.1188343	-1.64	0.100	.5767974	1.049607
pyperstudb~k		.8570974	.1366572	-0.97	0.333	.6270657	1.171514
pyperstudl~o		.7843441	.1196275	-1.59	0.111	.5816772	1.057624
pyperstudm~e		1.009285	.011461	0.81	0.416	.9870695	1.032
	male	.9911541	.3456375	-0.03	0.980	.5003895	1.963243
	age	.9971862	.0230154	-0.12	0.903	.9530819	1.043331
	Asian	(omitted)					
	Black	2.654166	2.955085	0.88	0.381	.2993785	23.53073
	Latino	4.197584	2.043197	2.95	0.003	1.61686	10.89749
	Other	8.93e-06	.0039272	-0.03	0.979	0	.
	AsianMale	(omitted)					
	BlackMale	.2527312	.4300425	-0.81	0.419	.0090006	7.096544
	LatinoMale	.7279074	.4695263	-0.49	0.622	.2055988	2.577102
	OtherMale	5.61e+11	7.04e+14	0.02	0.983	0	.
3							
	tenure	1.018262	.0176117	1.05	0.295	.9843224	1.053372
	exper	1.038711	.0254753	1.55	0.121	.9899618	1.089862
pyperstudlep		1.002888	.0161869	0.18	0.858	.9716591	1.035121
pyperstude~s		1.036654	.0125162	2.98	0.003	1.01241	1.061477
pyperstudw~e		1.064441	.1744348	0.38	0.703	.772025	1.467614
pyperstudb~k		1.073819	.184718	0.41	0.679	.7664916	1.504369
pyperstudl~o		1.034121	.1691568	0.21	0.837	.7504742	1.424974
pyperstudm~e		1.002747	.0113296	0.24	0.808	.9807856	1.0252
	male	1.831387	.5203103	2.13	0.033	1.049416	3.196043
	age	.9628247	.0220383	-1.66	0.098	.9205849	1.007002
	Asian	(omitted)					
	Black	2.459837	2.407289	0.92	0.358	.3613173	16.7465
	Latino	2.046116	.9503714	1.54	0.123	.823318	5.085021
	Other	1.611978	2.582797	0.30	0.766	.069746	37.25626
	AsianMale	(omitted)					
	BlackMale	.7425833	1.017036	-0.22	0.828	.0506928	10.87788
	LatinoMale	.4379331	.2716843	-1.33	0.183	.1298197	1.477321
	OtherMale	.5730888	1051.176	-0.00	1.000	0	.

-----  
-----> region = 19

Iteration 0: log likelihood = -820.33352  
Iteration 1: log likelihood = -754.46344  
Iteration 2: log likelihood = -752.40515  
Iteration 3: log likelihood = -752.27692  
Iteration 4: log likelihood = -752.26297  
Iteration 5: log likelihood = -752.26088  
Iteration 6: log likelihood = -752.2605  
Iteration 7: log likelihood = -752.26042  
Iteration 8: log likelihood = -752.2604

Multinomial logistic regression	Number of obs	=	759
	LR chi2(36)	=	136.15
	Prob > chi2	=	0.0000
Log likelihood = -752.2604	Pseudo R2	=	0.0830

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
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1		(base outcome)					
2							
	tenure	1.040046	.0138083	2.96	0.003	1.013331	1.067465
	exper	1.024606	.0184028	1.35	0.176	.9891643	1.061317
	pyperstudlep	1.016097	.0058458	2.78	0.006	1.004704	1.027619
	pyperstude~s	.9893414	.0093426	-1.13	0.256	.9711986	1.007823
	pyperstudw~e	.9185062	.0862351	-0.91	0.365	.7641278	1.104074
	pyperstudb~k	.9718579	.101814	-0.27	0.785	.79146	1.193374
	pyperstudl~o	.9325544	.0839352	-0.78	0.438	.781738	1.112467
	pyperstudm~e	.9907235	.0069673	-1.33	0.185	.9771615	1.004474
	male	1.806708	.5271594	2.03	0.043	1.019824	3.200742
	age	.9773198	.0148482	-1.51	0.131	.948647	1.006859
	Asian	2637716	4.20e+09	0.01	0.993	0	.
	Black	1.278046	.9061434	0.35	0.729	.3184482	5.129254
	Latino	1.462617	.378235	1.47	0.141	.8810661	2.428022
	Other	2.329291	3.400249	0.58	0.562	.1332471	40.7183
	AsianMale	8.69e-08	.000337	-0.00	0.997	0	.
	BlackMale	.8754978	.9032472	-0.13	0.897	.1158975	6.613569
	LatinoMale	.7223895	.2651661	-0.89	0.376	.3518216	1.483271
	OtherMale	.8273061	1.715311	-0.09	0.927	.0142171	48.14173

3							
	tenure	1.05019	.0146728	3.51	0.000	1.021822	1.079346
	exper	1.083481	.0229543	3.78	0.000	1.039413	1.129418
	pyperstudlep	1.026313	.0067101	3.97	0.000	1.013246	1.039549
	pyperstude~s	1.001446	.0109942	0.13	0.895	.9801277	1.023228
	pyperstudw~e	.9273707	.1016351	-0.69	0.491	.748111	1.149584
	pyperstudb~k	1.025057	.1238583	0.20	0.838	.8089036	1.29897
	pyperstudl~o	.9367999	.0975013	-0.63	0.530	.7639316	1.148786
	pyperstudm~e	.9958431	.0070046	-0.59	0.554	.9822085	1.009667
	male	1.375611	.4471515	0.98	0.327	.7274559	2.601263
	age	.9792974	.0184943	-1.11	0.268	.9437119	1.016225
	Asian	.4275916	1097.47	-0.00	1.000	0	.
	Black	.745205	.661748	-0.33	0.741	.130738	4.247659
	Latino	1.127871	.3188255	0.43	0.670	.6481007	1.962801
	Other	3.67e-06	.0032294	-0.01	0.989	0	.
	AsianMale	1470073	5.17e+09	0.00	0.997	0	.
	BlackMale	3.238991	3.783511	1.01	0.314	.3281779	31.96761
	LatinoMale	.6510554	.2746364	-1.02	0.309	.2848088	1.488273
	OtherMale	1.555994	1898.319	0.00	1.000	0	.

-----> region = 20

Iteration 0: log likelihood = -2194.6547  
Iteration 1: log likelihood = -2076.0297  
Iteration 2: log likelihood = -2072.3685  
Iteration 3: log likelihood = -2072.0403  
Iteration 4: log likelihood = -2071.9628  
Iteration 5: log likelihood = -2071.9477  
Iteration 6: log likelihood = -2071.9451  
Iteration 7: log likelihood = -2071.9445  
Iteration 8: log likelihood = -2071.9444  
Iteration 9: log likelihood = -2071.9444

Multinomial logistic regression                      Number of obs    =       2157  
   LR chi2(36)       =       245.42  
   Prob > chi2       =       0.0000  
Log likelihood = -2071.9444                           Pseudo R2        =       0.0559

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

```

-----+-----
2      |
      tenure | 1.003106 .0080129 0.39 0.698 .9875228 1.018934
      exper  | 1.003659 .0102166 0.36 0.720 .9838334 1.023884
pyperstudlep | .9871041 .0073578 -1.74 0.082 .9727879 1.001631
pyperstude~s | 1.006018 .0047679 1.27 0.206 .9967161 1.015406
pyperstudw~e | .9891658 .0426228 -0.25 0.800 .909057 1.076334
pyperstudb~k | .9781763 .0441749 -0.49 0.625 .8953163 1.068705
pyperstudl~o | .9822347 .0425042 -0.41 0.679 .9023629 1.069176
pyperstudm~e | .9981993 .0039988 -0.45 0.653 .9903924 1.006068
      male   | 1.58095 .2274386 3.18 0.001 1.192511 2.095915
      age    | .9621061 .0084061 -4.42 0.000 .9457707 .9787237
      Asian  | .3797772 .4202899 -0.87 0.382 .0434036 3.323015
      Black  | 1.728924 .4996148 1.89 0.058 .9812974 3.046151
      Latino | 1.331097 .2126349 1.79 0.073 .9732748 1.820473
      Other  | 2.540834 2.131639 1.11 0.266 .4907481 13.15509
      AsianMale | 5.31e+07 2.10e+11 0.00 0.996 0 .
      BlackMale | .9490566 .4228584 -0.12 0.907 .3963087 2.272744
      LatinoMale | 1.187803 .2784153 0.73 0.463 .7502862 1.880451
      OtherMale | 1.44e-07 .0002175 -0.01 0.992 0 .
-----+-----

```

```

3      |
      tenure | 1.017406 .0078331 2.24 0.025 1.002169 1.032875
      exper  | 1.065453 .0126382 5.34 0.000 1.040969 1.090514
pyperstudlep | 1.023945 .0067331 3.60 0.000 1.010833 1.037227
pyperstude~s | 1.023718 .0055303 4.34 0.000 1.012936 1.034615
pyperstudw~e | 1.086413 .0532467 1.69 0.091 .9869071 1.195951
pyperstudb~k | 1.044012 .0539326 0.83 0.404 .9434811 1.155254
pyperstudl~o | 1.049364 .0519628 0.97 0.331 .9523052 1.156316
pyperstudm~e | 1.009318 .003892 2.41 0.016 1.001719 1.016975
      male   | 1.512143 .234034 2.67 0.008 1.116484 2.048015
      age    | .9581479 .0103984 -3.94 0.000 .9379827 .9787466
      Asian  | 4.20e-07 .0004978 -0.01 0.990 0 .
      Black  | 1.595387 .4890556 1.52 0.128 .8748547 2.909353
      Latino | 1.145728 .1895822 0.82 0.411 .8283891 1.584633
      Other  | 7.70e-07 .0009419 -0.01 0.991 0 .
      AsianMale | 2.775308 2.06e+10 0.00 0.998 0 .
      BlackMale | .8679664 .4414368 -0.28 0.781 .3203263 2.35187
      LatinoMale | .9584563 .2448772 -0.17 0.868 .5808951 1.581419
      OtherMale | 4200690 5.14e+09 0.01 0.990 0 .
-----+-----

```

```

by pyschooltype:
. bysort pyschooltype: mlogit admin tenure exper pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile male age
> Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if
validcert2003a==1, rrr iter(20)

```

```

-----
-----> pyschooltype =

Iteration 0: log likelihood = -552.33627
Iteration 1: log likelihood = -488.41841
Iteration 2: log likelihood = -486.78646
Iteration 3: log likelihood = -486.68821
Iteration 4: log likelihood = -486.66589
Iteration 5: log likelihood = -486.66132
Iteration 6: log likelihood = -486.66058
Iteration 7: log likelihood = -486.6604
Iteration 8: log likelihood = -486.66036
Iteration 9: log likelihood = -486.66035

```

```

Multinomial logistic regression      Number of obs   =      516
                                      LR chi2(36)        =     131.35
                                      Prob > chi2         =     0.0000
Log likelihood = -486.66035          Pseudo R2        =     0.1189
-----

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9976778	.017671	-0.13	0.896	.9636375	1.03292
exper	1.040833	.0250525	1.66	0.096	.9928707	1.091111
pyperstudlep	.9948479	.0100008	-0.51	0.607	.9754386	1.014643
pyperstude~s	.9972728	.0097909	-0.28	0.781	.9782664	1.016649
pyperstudw~e	.9675624	.0258231	-1.24	0.217	.918251	1.019522
pyperstudb~k	.949902	.0278239	-1.75	0.079	.8969041	1.006032
pyperstudl~o	.9720498	.0269324	-1.02	0.306	.920671	1.026296
pyperstudm~e	.9572366	.0170059	-2.46	0.014	.9244793	.9911546
male	1.595649	.5019943	1.49	0.137	.8612863	2.956154
age	.927454	.0196881	-3.55	0.000	.8896577	.9668561
Asian	1.27e-06	.0010198	-0.02	0.987	0	.
Black	2.059483	.8794491	1.69	0.091	.8918149	4.756
Latino	.9957879	.4247187	-0.01	0.992	.4316347	2.297298
Other	.8046872	1.185598	-0.15	0.883	.0448226	14.44633
AsianMale	5.13e+13	1.39e+17	0.01	0.991	0	.
BlackMale	1.01704	.7353624	0.02	0.981	.2465379	4.195583
LatinoMale	1.468232	.8986422	0.63	0.530	.4423951	4.872804
OtherMale	5350740	1.39e+10	0.01	0.995	0	.
3						
tenure	1.017615	.0159543	1.11	0.265	.9868205	1.04937
exper	1.058228	.0233182	2.57	0.010	1.013498	1.104932
pyperstudlep	1.015499	.0108448	1.44	0.150	.9944642	1.036978
pyperstude~s	1.019616	.0095818	2.07	0.039	1.001008	1.03857
pyperstudw~e	1.011757	.0292756	0.40	0.686	.9559749	1.070795
pyperstudb~k	.9637974	.0309436	-1.15	0.251	.9050179	1.026395
pyperstudl~o	.9793978	.0292924	-0.70	0.486	.9236361	1.038526
pyperstudm~e	.9839357	.0105585	-1.51	0.131	.9634575	1.004849
male	2.095865	.6199999	2.50	0.012	1.173709	3.742538
age	.9803881	.0200457	-0.97	0.333	.9418761	1.020475
Asian	.7743059	.9378752	-0.21	0.833	.0720938	8.316245
Black	1.489716	.6968596	0.85	0.394	.5955641	3.726307
Latino	.4400905	.222586	-1.62	0.105	.1633154	1.185924
Other	.9770188	1.482271	-0.02	0.988	.0499475	19.11138
AsianMale	3.797635	15861.66	0.00	1.000	0	.
BlackMale	1.174943	.8698024	0.22	0.828	.2753456	5.013668
LatinoMale	3.814571	2.467145	2.07	0.038	1.07378	13.55115
OtherMale	1.232389	5147.347	0.00	1.000	0	.

```

-----
-----> pyschooltype = B

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -625.06896
Iteration 1: log likelihood = -587.47541
Iteration 2: log likelihood = -586.69935
Iteration 3: log likelihood = -582.07314
Iteration 4: log likelihood = -580.8659
Iteration 5: log likelihood = -580.66019
Iteration 6: log likelihood = -580.64665
Iteration 7: log likelihood = -580.64436
Iteration 8: log likelihood = -580.64383
Iteration 9: log likelihood = -580.64371
Iteration 10: log likelihood = -580.64368 (not concave)
Iteration 11: log likelihood = -580.64368 (not concave)
Iteration 12: log likelihood = -580.64368 (not concave)
Iteration 13: log likelihood = -580.64368 (not concave)
Iteration 14: log likelihood = -580.64368 (not concave)
Iteration 15: log likelihood = -580.64368 (not concave)
Iteration 16: log likelihood = -580.64368 (not concave)

```

```

Iteration 17: log likelihood = -580.64368 (not concave)
Iteration 18: log likelihood = -580.64368 (not concave)
Iteration 19: log likelihood = -580.64368 (not concave)
Iteration 20: log likelihood = -580.64368 (not concave)
convergence not achieved

```

```

Multinomial logistic regression      Number of obs   =      660
                                     LR chi2(31)        =      88.85
                                     Prob > chi2         =      0.0000
Log likelihood = -580.64368          Pseudo R2        =      0.0711

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.073532	.0217983	3.49	0.000	1.031647	1.117117
	exper	.9573089	.0245026	-1.70	0.088	.9104693	1.006558
pypersistudlep		1.01597	.0099807	1.61	0.107	.9965958	1.035722
pypersistude~s		.9821088	.0073158	-2.42	0.015	.9678742	.9965527
pypersistudw~e		.9749185	.0320295	-0.77	0.439	.9141204	1.03976
pypersistudb~k		1.009623	.0345776	0.28	0.780	.944077	1.07972
pypersistudl~o		.9936591	.0323725	-0.20	0.845	.9321935	1.059177
pypersistudm~e		.9917114	.0045556	-1.81	0.070	.9828227	1.00068
	male	1.051356	.3441776	0.15	0.878	.5534727	1.997118
	age	.9690124	.0196216	-1.55	0.120	.9313078	1.008243
	Asian	(omitted)					
	Black	2.590095	1.441068	1.71	0.087	.8704162	7.707335
	Latino	2.245433	1.279149	1.42	0.156	.7351887	6.858062
	Other	22.07058	29.71405	2.30	0.022	1.576974	308.8893
	AsianMale	(omitted)					
	BlackMale	.4600037	.3507108	-1.02	0.308	.1032281	2.049863
	LatinoMale	1.502603	1.086813	0.56	0.573	.3640669	6.201649
	OtherMale	1.18e-13	7.60e-08	-0.00	1.000	0	.
3							
	tenure	.9793196	.0114865	-1.78	0.075	.9570633	1.002093
	exper	1.043035	.0162171	2.71	0.007	1.01173	1.075309
pypersistudlep		.997704	.0083787	-0.27	0.784	.9814165	1.014262
pypersistude~s		.997913	.0048924	-0.43	0.670	.98837	1.007548
pypersistudw~e		1.035101	.0326088	1.10	0.273	.9731223	1.101028
pypersistudb~k		1.042562	.0338743	1.28	0.200	.9782395	1.111114
pypersistudl~o		1.038388	.0325765	1.20	0.230	.9764631	1.104241
pypersistudm~e		1.00279	.0031493	0.89	0.375	.9966363	1.008981
	male	1.206932	.2404592	0.94	0.345	.816764	1.783484
	age	.9744743	.0141552	-1.78	0.075	.9471219	1.002617
	Asian	(omitted)					
	Black	.8947351	.459511	-0.22	0.829	.326995	2.448206
	Latino	.8330744	.4169992	-0.36	0.715	.3123281	2.222064
	Other	3.88e-10	3.63e-10	-23.21	0.000	6.23e-11	2.42e-09
	AsianMale	(omitted)					
	BlackMale	1.192559	.7773767	0.27	0.787	.3323682	4.278984
	LatinoMale	1.54324	.9810836	0.68	0.495	.4439147	5.364974
	OtherMale	2.83e+09	.	.	.	.	.

Warning: convergence not achieved

```

-----
-----> pyschooltype = E

```

```

Iteration 0: log likelihood = -10150.243
Iteration 1: log likelihood = -9626.7773
Iteration 2: log likelihood = -9622.5665
Iteration 3: log likelihood = -9622.5649
Iteration 4: log likelihood = -9622.5649

```

Multinomial logistic regression

Number of obs = 9393

LR chi2(36) = 1055.36

Prob > chi2 = 0.0000

Pseudo R2 = 0.0520

Log likelihood = -9622.5649

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.002209	.0042712	0.52	0.605	.9938723	1.010615
exper	1.004213	.0056707	0.74	0.457	.9931599	1.015389
pyperstudlep	1.009775	.0020447	4.80	0.000	1.005775	1.01379
pyperstude~s	.9835398	.0022904	-7.13	0.000	.9790608	.9880392
pyperstudw~e	.9951741	.0060237	-0.80	0.424	.9834375	1.007051
pyperstudb~k	1.002236	.0067253	0.33	0.739	.9891413	1.015505
pyperstudl~o	1.003593	.0063998	0.56	0.574	.9911274	1.016215
pyperstudm~e	1.009692	.0038875	2.51	0.012	1.002101	1.01734
male	1.686369	.1509983	5.84	0.000	1.414932	2.009878
age	.974847	.0045001	-5.52	0.000	.9660668	.983707
Asian	.8004687	.4520799	-0.39	0.694	.2646136	2.421456
Black	1.348015	.1372185	2.93	0.003	1.104202	1.645664
Latino	1.031705	.090288	0.36	0.721	.8690885	1.224749
Other	.7804701	.2550359	-0.76	0.448	.4113461	1.48083
AsianMale	1.903494	1.648878	0.74	0.457	.3485048	10.39667
BlackMale	.8925574	.1833487	-0.55	0.580	.5967367	1.335026
LatinoMale	.930239	.1454095	-0.46	0.644	.6847621	1.263716
OtherMale	1.526084	1.101017	0.59	0.558	.3710785	6.276114
3						
tenure	1.017714	.0036124	4.95	0.000	1.010658	1.024819
exper	1.059262	.005546	11.00	0.000	1.048448	1.070188
pyperstudlep	1.000822	.0018929	0.43	0.664	.9971191	1.004539
pyperstude~s	1.00229	.0020877	1.10	0.272	.9982062	1.00639
pyperstudw~e	1.012457	.006192	2.02	0.043	1.000393	1.024666
pyperstudb~k	1.001108	.006746	0.16	0.869	.9879733	1.014418
pyperstudl~o	1.002782	.0064372	0.43	0.665	.9902445	1.015478
pyperstudm~e	1.005351	.0036202	1.48	0.138	.9982811	1.012472
male	2.650636	.2057387	12.56	0.000	2.27657	3.086165
age	.9805016	.0044198	-4.37	0.000	.9718771	.9892026
Asian	1.241115	.6672181	0.40	0.688	.4327227	3.559708
Black	1.036719	.104744	0.36	0.721	.8504731	1.263751
Latino	1.256897	.1025365	2.80	0.005	1.071173	1.474823
Other	.4816568	.1729991	-2.03	0.042	.2382359	.9737963
AsianMale	.5860212	.5182794	-0.60	0.546	.1035398	3.316799
BlackMale	.6910189	.1369373	-1.87	0.062	.4686079	1.018991
LatinoMale	.5565519	.0815107	-4.00	0.000	.417678	.7416
OtherMale	.8535108	.6466734	-0.21	0.834	.1933238	3.76819

-----> pyschooltype = M

Iteration 0: log likelihood = -5736.134  
 Iteration 1: log likelihood = -5458.5735  
 Iteration 2: log likelihood = -5454.62  
 Iteration 3: log likelihood = -5454.6068  
 Iteration 4: log likelihood = -5454.6068

Multinomial logistic regression

Number of obs = 5445

LR chi2(36) = 563.05

Prob > chi2 = 0.0000

Pseudo R2 = 0.0491

Log likelihood = -5454.6068

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-------	-----	-----------	---	------	----------------------	--

1		(base outcome)					
2							
	tenure	1.016208	.0050488	3.24	0.001	1.00636	1.026152
	exper	1.001085	.0062615	0.17	0.862	.988888	1.013433
	pyperstudlep	1.013737	.0040638	3.40	0.001	1.005803	1.021733
	pyperstude~s	.9979192	.0028575	-0.73	0.467	.9923342	1.003536
	pyperstudw~e	.9850772	.0084603	-1.75	0.080	.9686341	1.001799
	pyperstudb~k	.9884053	.0094063	-1.23	0.220	.9701402	1.007014
	pyperstudl~o	.9819135	.0089869	-1.99	0.046	.9644566	.9996863
	pyperstudm~e	1.004227	.0036086	1.17	0.241	.9971787	1.011324
	male	1.478708	.119436	4.84	0.000	1.262206	1.732345
	age	.9608695	.0050256	-7.63	0.000	.9510698	.9707702
	Asian	.5869132	.3595369	-0.87	0.384	.1766593	1.949894
	Black	1.056619	.1336316	0.44	0.663	.8246434	1.35385
	Latino	1.751245	.2196636	4.47	0.000	1.369551	2.239317
	Other	.5251825	.2704428	-1.25	0.211	.1914185	1.440909
	AsianMale	1.774587	1.59503	0.64	0.523	.3048051	10.33171
	BlackMale	1.407841	.2667243	1.81	0.071	.9711541	2.040888
	LatinoMale	1.02593	.1707219	0.15	0.878	.7404095	1.421555
	OtherMale	1.667481	1.089082	0.78	0.434	.4635705	5.997997
3							
	tenure	1.011998	.0050042	2.41	0.016	1.002238	1.021854
	exper	1.06328	.0078362	8.33	0.000	1.048032	1.07875
	pyperstudlep	1.006463	.0049967	1.30	0.194	.9967169	1.016304
	pyperstude~s	1.009298	.0032915	2.84	0.005	1.002867	1.01577
	pyperstudw~e	1.02505	.0116968	2.17	0.030	1.002379	1.048234
	pyperstudb~k	1.007975	.0126425	0.63	0.527	.9834977	1.033061
	pyperstudl~o	1.007741	.012161	0.64	0.523	.9841852	1.03186
	pyperstudm~e	1.003426	.0042562	0.81	0.420	.9951182	1.011802
	male	1.88974	.1669749	7.20	0.000	1.589246	2.247052
	age	.9607778	.0064356	-5.97	0.000	.9482467	.9734744
	Asian	.340599	.3632808	-1.01	0.313	.0421073	2.755051
	Black	.8866603	.146815	-0.73	0.468	.6409345	1.226594
	Latino	1.499778	.2320526	2.62	0.009	1.107452	2.031088
	Other	.3570995	.2699108	-1.36	0.173	.0811737	1.570952
	AsianMale	5.13516	6.411163	1.31	0.190	.4444704	59.32874
	BlackMale	1.774137	.4009613	2.54	0.011	1.139233	2.762879
	LatinoMale	.8315456	.1649107	-0.93	0.352	.563738	1.226577
	OtherMale	2.348717	2.059553	0.97	0.330	.4211389	13.09894

-----> pyschooltype = S

Iteration 0: log likelihood = -6922.5073  
Iteration 1: log likelihood = -6631.1211  
Iteration 2: log likelihood = -6618.7856  
Iteration 3: log likelihood = -6618.6948  
Iteration 4: log likelihood = -6618.6948

Multinomial logistic regression	Number of obs	=	7092
	LR chi2(36)	=	607.62
	Prob > chi2	=	0.0000
Log likelihood = -6618.6948	Pseudo R2	=	0.0439

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005852	.0038507	1.52	0.127	.9983327	1.013427
	exper	1.000476	.0051748	0.09	0.927	.9903846	1.01067
	pyperstudlep	1.020022	.0043396	4.66	0.000	1.011552	1.028563



pyperstude~s		.9897451	.002228	-4.58	0.000	.985388	.9941215
pyperstudw~e		.9996238	.0069013	-0.05	0.957	.9861886	1.013242
pyperstudb~k		1.006404	.0074885	0.86	0.391	.9918331	1.021189
pyperstudl~o		1.004409	.0072892	0.61	0.544	.9902235	1.018798
pyperstudm~e		.9985272	.0021643	-0.68	0.496	.9942943	1.002778
male		1.140732	.0775389	1.94	0.053	.9984465	1.303293
age		.9726016	.0045489	-5.94	0.000	.9637267	.9815584
Asian		.4646841	.3008542	-1.18	0.237	.1306344	1.652944
Black		1.300309	.1603631	2.13	0.033	1.021105	1.655856
Latino		1.511927	.1832601	3.41	0.001	1.192221	1.917364
Other		.5830309	.3003124	-1.05	0.295	.2124455	1.600057
AsianMale		2.134756	1.940155	0.83	0.404	.3595346	12.67523
BlackMale		1.928771	.3255626	3.89	0.000	1.385492	2.685081
LatinoMale		1.188739	.1767508	1.16	0.245	.8882254	1.590924
OtherMale		2.146265	1.386269	1.18	0.237	.6051926	7.61155
-----							
3							
tenure		.9985751	.0044319	-0.32	0.748	.9899263	1.007299
exper		1.044024	.0070179	6.41	0.000	1.030359	1.057869
pyperstudlep		1.003208	.0061489	0.52	0.601	.991228	1.015332
pyperstude~s		1.012298	.002713	4.56	0.000	1.006994	1.017629
pyperstudw~e		1.064779	.0137806	4.85	0.000	1.038109	1.092134
pyperstudb~k		1.051043	.0142984	3.66	0.000	1.023389	1.079445
pyperstudl~o		1.04856	.0139506	3.56	0.000	1.021571	1.076263
pyperstudm~e		1.01575	.0021111	7.52	0.000	1.01162	1.019896
male		1.894778	.168579	7.18	0.000	1.591574	2.255745
age		.9756543	.0063046	-3.81	0.000	.9633755	.9880897
Asian		.8572909	.6652222	-0.20	0.843	.1873403	3.923063
Black		.7560049	.1602998	-1.32	0.187	.4989302	1.145538
Latino		1.475099	.2541351	2.26	0.024	1.052382	2.067612
Other		1.458599	.8305375	0.66	0.507	.47781	4.45263
AsianMale		.6733969	.8943927	-0.30	0.766	.0498565	9.095372
BlackMale		1.924821	.5056473	2.49	0.013	1.150221	3.221066
LatinoMale		.7021067	.1442903	-1.72	0.085	.4693235	1.05035
OtherMale		1.198522	.8523291	0.25	0.799	.2973817	4.830344
-----							

## 2003-04

```
. mlogit admin tenure exper acctrating2 pyperstudlep pyperstudecodis pyperstudwhite
pyperstudblack pyperstudlatino pyperstudmobile ma
> le age Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if
validcert2004a==1, rrr
```

```
Iteration 0: log likelihood = -24189.362
Iteration 1: log likelihood = -23045.246
Iteration 2: log likelihood = -23024.356
Iteration 3: log likelihood = -23024.322
Iteration 4: log likelihood = -23024.322
```

```
Multinomial logistic regression      Number of obs   =      22835
                                      LR chi2(38)        =      2330.08
                                      Prob > chi2         =      0.0000
Log likelihood = -23024.322          Pseudo R2        =      0.0482
```

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.009167	.0024851	3.71	0.000	1.004308 1.01405
exper		1.006603	.0032125	2.06	0.039	1.000327 1.012919
acctrating2		.9579374	.0284362	-1.45	0.148	.9037939 1.015325
pyperstudlep		1.009129	.0013505	6.79	0.000	1.006485 1.011779
pyperstude~s		.9891095	.0012683	-8.54	0.000	.9866267 .9915985
pyperstudw~e		.9893666	.0036456	-2.90	0.004	.9822471 .9965378

pyperstudb~k		.9938998	.0040753	-1.49	0.136	.9859443	1.001919
pyperstudl~o		.994843	.0038843	-1.32	0.185	.9872589	1.002485
pyperstudm~e		1.008493	.0021656	3.94	0.000	1.004258	1.012747
male		1.191842	.0495425	4.22	0.000	1.098591	1.293009
age		.9718868	.0026245	-10.56	0.000	.9667564	.9770445
Asian		.6461345	.1920707	-1.47	0.142	.3608226	1.15705
Black		1.180773	.0763507	2.57	0.010	1.040223	1.340314
Latino		1.282652	.075526	4.23	0.000	1.142846	1.43956
Other		.7233674	.1604613	-1.46	0.144	.4683179	1.117319
AsianMale		1.910634	.8543886	1.45	0.148	.7953188	4.590013
BlackMale		1.761808	.1800526	5.54	0.000	1.442009	2.152531
LatinoMale		1.270611	.1067103	2.85	0.004	1.077769	1.497957
OtherMale		1.914336	.6257183	1.99	0.047	1.008777	3.632798

3							
tenure		1.01384	.0023467	5.94	0.000	1.009251	1.01845
exper		1.057982	.0036443	16.36	0.000	1.050863	1.065148
acctrating2		1.422751	.0417263	12.02	0.000	1.343275	1.506929
pyperstudlep		1.012204	.0014155	8.67	0.000	1.009434	1.014982
pyperstude~s		1.019792	.0013619	14.67	0.000	1.017126	1.022465
pyperstudw~e		1.013272	.0043808	3.05	0.002	1.004722	1.021895
pyperstudb~k		.9982093	.0047729	-0.37	0.708	.9888984	1.007608
pyperstudl~o		.9929193	.0045265	-1.56	0.119	.9840871	1.001831
pyperstudm~e		.9931635	.0025541	-2.67	0.008	.9881702	.9981819
male		1.331311	.0561951	6.78	0.000	1.225603	1.446135
age		.9757221	.0030093	-7.97	0.000	.9698419	.9816381
Asian		.7446488	.2582289	-0.85	0.395	.3773743	1.469369
Black		.7894031	.0595793	-3.13	0.002	.6808561	.9152553
Latino		1.173781	.0748697	2.51	0.012	1.035841	1.33009
Other		.4983354	.1349681	-2.57	0.010	.2930787	.8473431
AsianMale		.8019375	.4574871	-0.39	0.699	.2621497	2.453193
BlackMale		1.737414	.2054541	4.67	0.000	1.377991	2.190586
LatinoMale		1.015916	.094044	0.17	0.865	.8473476	1.218019
OtherMale		1.901416	.7624886	1.60	0.109	.8664354	4.172711

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=22835)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		38.156	2	0.000
exper		293.827	2	0.000
acctrating2		180.359	2	0.000
pyperstudlep		89.956	2	0.000
pyperstude~s		428.309	2	0.000
pyperstudw~e		27.650	2	0.000
pyperstudb~k		2.242	2	0.326
pyperstudl~o		3.128	2	0.209
pyperstudm~e		33.492	2	0.000
male		50.340	2	0.000
age		140.908	2	0.000
Asian		2.514	2	0.285
Black		23.998	2	0.000
Latino		19.068	2	0.000
Other		8.036	2	0.018
AsianMale		2.742	2	0.254
BlackMale		38.748	2	0.000
LatinoMale		9.153	2	0.010
OtherMale		5.171	2	0.075

\*\*\*\* Wald tests for independent variables (N=22835)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	38.075	2	0.000
exper	276.409	2	0.000
acctrating2	181.619	2	0.000
pyperstudlep	88.733	2	0.000
pyperstude~s	413.691	2	0.000
pyperstudw~e	27.239	2	0.000
pyperstudb~k	2.250	2	0.325
pyperstudl~o	3.126	2	0.210
pyperstudm~e	33.204	2	0.000
male	50.265	2	0.000
age	138.385	2	0.000
Asian	2.437	2	0.296
Black	23.726	2	0.000
Latino	19.075	2	0.000
Other	7.471	2	0.024
AsianMale	2.742	2	0.254
BlackMale	38.026	2	0.000
LatinoMale	9.162	2	0.010
OtherMale	5.112	2	0.078

\*\*\*\* Hausman tests of IIA assumption (N=22835)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	111.539	19	0.000	against Ho
3	-7.784	19	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=22835)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-4981.981	-4973.719	16.524	20	0.684	for Ho
3	-5451.061	-5444.298	13.526	20	0.854	for Ho

\*\*\*\* Wald tests for combining alternatives (N=22835)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2-	3   1499.117	19	0.000
2-	1   533.963	19	0.000
3-	1   1287.871	19	0.000

\*\*\*\* LR tests for combining alternatives (N=22835)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2-	3   1681.730	19	0.000

```

2-      1 | 554.168  19  0.000
3-      1 | 1423.244 19  0.000
-----
by region:
. bysort region: mlogit admin tenure exper acctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyp
> erstudmobile male age Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale
if validcert2004a==1, rrr iter(20)

```

```

-----> region = 1

```

```

note: OtherMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -1725.4395
Iteration 1: log likelihood = -1635.8432
Iteration 2: log likelihood = -1632.8416
Iteration 3: log likelihood = -1632.5298
Iteration 4: log likelihood = -1632.4886
Iteration 5: log likelihood = -1632.4796
Iteration 6: log likelihood = -1632.4774
Iteration 7: log likelihood = -1632.477
Iteration 8: log likelihood = -1632.4769
Iteration 9: log likelihood = -1632.4769

```

```

Multinomial logistic regression      Number of obs   =      1602
                                      LR chi2(36)        =      185.93
                                      Prob > chi2         =      0.0000
Log likelihood = -1632.4769          Pseudo R2        =      0.0539

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.022024	.0093906	2.37	0.018	1.003784	1.040596
	exper	1.001756	.0135897	0.13	0.897	.9754721	1.028749
	acctrating2	.8076145	.1028066	-1.68	0.093	.6292874	1.036476
	pyperstudlep	1.007838	.0031451	2.50	0.012	1.001692	1.014021
	pyperstude~s	.9981838	.0062063	-0.29	0.770	.9860935	1.010422
	pyperstudw~e	1.013092	.0839378	0.16	0.875	.8612398	1.191718
	pyperstudb~k	.9371971	.1215723	-0.50	0.617	.7267982	1.208504
	pyperstudl~o	1.000802	.0756964	0.01	0.992	.8629123	1.160725
	pyperstudm~e	.9897164	.008229	-1.24	0.214	.9737186	1.005977
	male	1.334747	.3766863	1.02	0.306	.7676741	2.320712
	age	.9677146	.0109022	-2.91	0.004	.9465809	.9893201
	Asian	2.37e-07	.000446	-0.01	0.994	0	.
	Black	4.32e-07	.0008135	-0.01	0.994	0	.
	Latino	1.037658	.2130858	0.18	0.857	.693837	1.551853
	Other	1.603213	1.636914	0.46	0.644	.2167187	11.86003
	AsianMale	2694189	5.07e+09	0.01	0.994	0	.
	BlackMale	1387097	2.61e+09	0.01	0.994	0	.
	LatinoMale	1.052426	.3290731	0.16	0.870	.5702135	1.942432
	OtherMale	(omitted)					
3							
	tenure	1.024357	.0095529	2.58	0.010	1.005803	1.043252
	exper	1.093309	.0187437	5.20	0.000	1.057183	1.13067
	acctrating2	1.231893	.16854	1.52	0.127	.942143	1.610754
	pyperstudlep	1.019558	.0035219	5.61	0.000	1.012679	1.026484
	pyperstude~s	.9982127	.0072957	-0.24	0.807	.9840153	1.012615
	pyperstudw~e	1.140483	.1148479	1.31	0.192	.9362066	1.389331
	pyperstudb~k	.9946812	.2693205	-0.02	0.984	.5850757	1.691047
	pyperstudl~o	1.115597	.102438	1.19	0.234	.9318525	1.335573
	pyperstudm~e	.9802481	.01043	-1.87	0.061	.9600174	1.000905
	male	.832507	.2815742	-0.54	0.588	.4290331	1.615418
	age	.9452741	.0144268	-3.69	0.000	.9174168	.9739773

Asian		3.83e-07	.0008515	-0.01	0.995	0	.
Black		9.19e-07	.0020438	-0.01	0.995	0	.
Latino		.8947332	.2000225	-0.50	0.619	.5773017	1.386706
Other		1.899635	1.992165	0.61	0.541	.2432244	14.83656
AsianMale		8.472975	20514.29	0.00	0.999	0	.
BlackMale		2.001829	4697.436	0.00	1.000	0	.
LatinoMale		1.589191	.592339	1.24	0.214	.7654345	3.299471
OtherMale		(omitted)					

-----> region = 2

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -762.95746
Iteration 1: log likelihood = -712.17598
Iteration 2: log likelihood = -709.95058
Iteration 3: log likelihood = -709.64909
Iteration 4: log likelihood = -709.59948
Iteration 5: log likelihood = -709.5883
Iteration 6: log likelihood = -709.58549
Iteration 7: log likelihood = -709.58496
Iteration 8: log likelihood = -709.58487
Iteration 9: log likelihood = -709.58485
```

Multinomial logistic regression	Number of obs	=	767
	LR chi2(36)	=	106.75
	Prob > chi2	=	0.0000
Log likelihood = -709.58485	Pseudo R2	=	0.0700

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.00554	.0154488	0.36	0.719	.9757121 1.03628
exper		.9953279	.0197085	-0.24	0.813	.95744 1.034715
acctrating2		.9532188	.1942502	-0.24	0.814	.63934 1.421194
pyperstudlep		.985889	.0133334	-1.05	0.293	.9600993 1.012372
pyperstude~s		1.005922	.0077216	0.77	0.442	.9909014 1.021171
pyperstudw~e		.9252384	.0806307	-0.89	0.373	.7799648 1.09757
pyperstudb~k		.9561243	.0881854	-0.49	0.627	.798006 1.145572
pyperstudl~o		.9236675	.0773006	-0.95	0.343	.7839343 1.088308
pyperstudm~e		1.011778	.0096654	1.23	0.220	.99301 1.0309
male		1.198747	.3470808	0.63	0.531	.6796319 2.114371
age		.9660118	.0154347	-2.16	0.030	.9362292 .9967419
Asian		4.30e-07	.0006909	-0.01	0.993	0 .
Black		1.296211	1.266242	0.27	0.791	.1910532 8.79422
Latino		1.043621	.2798736	0.16	0.874	.6169801 1.765282
Other		4.51e-07	.0004595	-0.01	0.989	0 .
AsianMale		(omitted)				
BlackMale		2.825339	3.806172	0.77	0.441	.2015439 39.60697
LatinoMale		1.629591	.6429406	1.24	0.216	.7520467 3.531121
OtherMale		6004582	6.12e+09	0.02	0.988	0 .
3						
tenure		1.002332	.0129282	0.18	0.857	.9773105 1.027994
exper		1.056745	.0204387	2.85	0.004	1.017436 1.097573
acctrating2		1.833249	.3335555	3.33	0.001	1.283357 2.61876
pyperstudlep		.9802964	.0150344	-1.30	0.194	.951268 1.010211
pyperstude~s		1.029137	.0084447	3.50	0.000	1.012718 1.045822
pyperstudw~e		1.088339	.098571	0.93	0.350	.9113202 1.299744
pyperstudb~k		1.06609	.1048495	0.65	0.515	.8791813 1.292734
pyperstudl~o		1.067589	.0933283	0.75	0.454	.8994812 1.267114
pyperstudm~e		.9844568	.0133323	-1.16	0.247	.9586697 1.010938
male		1.302763	.374704	0.92	0.358	.7413798 2.289234

age		.9724384	.0169243	-1.61	0.108	.9398268	1.006182
Asian		1.13e-06	.001771	-0.01	0.993	0	.
Black		3.981576	3.386972	1.62	0.104	.7515728	21.09303
Latino		1.046709	.2700787	0.18	0.860	.6312381	1.735635
Other		8.48e-07	.0007982	-0.01	0.988	0	.
AsianMale		(omitted)					
BlackMale		.6650212	.8863528	-0.31	0.760	.0487908	9.064272
LatinoMale		1.291721	.5058078	0.65	0.513	.5995955	2.782783
OtherMale		2.357502	3389.038	0.00	1.000	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -296.76628  
Iteration 1: log likelihood = -270.15374  
Iteration 2: log likelihood = -269.16576  
Iteration 3: log likelihood = -269.14892  
Iteration 4: log likelihood = -269.14562  
Iteration 5: log likelihood = -269.14491  
Iteration 6: log likelihood = -269.14475  
Iteration 7: log likelihood = -269.14471  
Iteration 8: log likelihood = -269.1447

Multinomial logistic regression	Number of obs	=	281
	LR chi2(32)	=	55.24
	Prob > chi2	=	0.0065
Log likelihood = -269.1447	Pseudo R2	=	0.0931

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.024976	.0258706	0.98	0.328	.9755043 1.076957
exper		.9424401	.0306314	-1.82	0.068	.884276 1.00443
acctrating2		1.26437	.3656151	0.81	0.417	.7173548 2.228508
pyperstudlep		.9627728	.0374399	-0.98	0.329	.8921187 1.039023
pyperstude~s		.9720102	.0184369	-1.50	0.134	.9365379 1.008826
pyperstudw~e		.9941396	.0554558	-0.11	0.916	.8911792 1.108995
pyperstudb~k		1.043648	.060215	0.74	0.459	.9320569 1.168599
pyperstudl~o		1.035044	.0623588	0.57	0.568	.9197631 1.164773
pyperstudm~e		.9706143	.028007	-1.03	0.301	.917245 1.027089
male		1.33242	.510623	0.75	0.454	.6286884 2.823884
age		1.001849	.0300698	0.06	0.951	.9446133 1.062553
Asian		(omitted)				
Black		4.633005	3.736127	1.90	0.057	.953763 22.50531
Latino		1.967326	1.274886	1.04	0.296	.552424 7.006162
Other		.6078162	1092.413	-0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		.5480934	.913931	-0.36	0.718	.0208689 14.39495
LatinoMale		.667084	.6471116	-0.42	0.676	.0996469 4.465779
OtherMale		(omitted)				
3						
tenure		1.016494	.0206771	0.80	0.421	.9767653 1.057839
exper		1.035055	.0331531	1.08	0.282	.9720732 1.102116
acctrating2		1.075667	.2686404	0.29	0.770	.6593202 1.754929
pyperstudlep		.9579126	.0281689	-1.46	0.144	.9042635 1.014745
pyperstude~s		1.054194	.0185838	2.99	0.003	1.018393 1.091254
pyperstudw~e		1.118698	.065671	1.91	0.056	.9971134 1.255107
pyperstudb~k		1.079238	.0653155	1.26	0.208	.9585231 1.215156
pyperstudl~o		1.089205	.0685781	1.36	0.175	.9627566 1.232261

```

pyperstudm~e | .9660608 .0289836 -1.15 0.250 .9108919 1.024571
      male | 1.161141 .4006676 0.43 0.665 .5904269 2.283514
      age | .9895607 .0306138 -0.34 0.734 .9313417 1.051419
      Asian | (omitted)
      Black | 2.697177 2.264178 1.18 0.237 .5204246 13.97851
      Latino | 1.613211 1.067792 0.72 0.470 .4408357 5.903443
      Other | 5151174 5.33e+09 0.01 0.988 0 .
      AsianMale | (omitted)
      BlackMale | .553415 .9240394 -0.35 0.723 .0209796 14.59841
      LatinoMale | .7742087 .7601961 -0.26 0.794 .1129959 5.304611
      OtherMale | (omitted)

```

-----> region = 4

```

Iteration 0: log likelihood = -4993.8732
Iteration 1: log likelihood = -4720.3197
Iteration 2: log likelihood = -4710.564
Iteration 3: log likelihood = -4710.364
Iteration 4: log likelihood = -4710.3184
Iteration 5: log likelihood = -4710.309
Iteration 6: log likelihood = -4710.307
Iteration 7: log likelihood = -4710.3065
Iteration 8: log likelihood = -4710.3064
Iteration 9: log likelihood = -4710.3064

```

```

Multinomial logistic regression      Number of obs   =      4855
                                      LR chi2(38)       =      567.13
                                      Prob > chi2       =      0.0000
Log likelihood = -4710.3064          Pseudo R2       =      0.0568

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.013706	.0052953	2.61	0.009	1.003381 1.024138	
exper	1.037964	.0072828	5.31	0.000	1.023787 1.052336	
acctrating2	1.019932	.0618231	0.33	0.745	.9056817 1.148594	
pyperstudlep	1.001307	.00321	0.41	0.684	.9950347 1.007618	
pyperstude~s	.9913793	.003409	-2.52	0.012	.9847204 .9980833	
pyperstudw~e	1.000556	.0057436	0.10	0.923	.9893614 1.011876	
pyperstudb~k	1.004463	.0070982	0.63	0.529	.9906465 1.018472	
pyperstudl~o	1.005793	.0068788	0.84	0.398	.9924007 1.019366	
pyperstudm~e	1.011435	.0055852	2.06	0.039	1.000547 1.022441	
male	1.01847	.0919381	0.20	0.839	.8533158 1.21559	
age	.9483976	.0057326	-8.77	0.000	.9372281 .9597001	
Asian	.8170373	.3618157	-0.46	0.648	.3430022 1.946197	
Black	.8107575	.0882015	-1.93	0.054	.6550728 1.003442	
Latino	1.226374	.1696152	1.48	0.140	.9351821 1.608234	
Other	.4304691	.2021384	-1.79	0.073	.1714894 1.080554	
AsianMale	1.442864	1.260663	0.42	0.675	.2603222 7.99723	
BlackMale	2.250562	.3844411	4.75	0.000	1.610233 3.145526	
LatinoMale	1.573798	.358769	1.99	0.047	1.006713 2.460323	
OtherMale	1.097046	1.020307	0.10	0.921	.1772416 6.790223	
3						
tenure	1.038917	.0060838	6.52	0.000	1.027061 1.05091	
exper	1.064393	.0092015	7.22	0.000	1.04651 1.082581	
acctrating2	1.39527	.0974066	4.77	0.000	1.216842 1.599861	
pyperstudlep	1.014542	.0038217	3.83	0.000	1.007079 1.02206	
pyperstude~s	1.020217	.0044009	4.64	0.000	1.011628 1.028879	
pyperstudw~e	1.013559	.0072593	1.88	0.060	.99943 1.027887	
pyperstudb~k	1.001593	.0089551	0.18	0.859	.9841943 1.0193	
pyperstudl~o	.9880218	.0084235	-1.41	0.158	.9716493 1.00467	

pyperstudm~e		.9984927	.0072485	-0.21	0.835	.9843865	1.012801
male		.9007387	.0991732	-0.95	0.342	.7259049	1.117681
age		.9667269	.0073911	-4.43	0.000	.9523486	.9813222
Asian		.5692699	.3695511	-0.87	0.385	.1594948	2.031842
Black		.4569816	.0643924	-5.56	0.000	.3467029	.6023376
Latino		1.171122	.1894021	0.98	0.329	.8529817	1.60792
Other		.4575322	.2577015	-1.39	0.165	.1517012	1.379922
AsianMale		4.76e-06	.0021539	-0.03	0.978	0	.
BlackMale		2.551046	.5646738	4.23	0.000	1.653122	3.936694
LatinoMale		1.496652	.4298597	1.40	0.160	.8524004	2.627834
OtherMale		3.056831	3.025857	1.13	0.259	.439233	21.27394

-----> region = 5

Iteration 0: log likelihood = -502.42187  
Iteration 1: log likelihood = -456.18518  
Iteration 2: log likelihood = -454.7094  
Iteration 3: log likelihood = -454.53919  
Iteration 4: log likelihood = -454.5086  
Iteration 5: log likelihood = -454.50191  
Iteration 6: log likelihood = -454.50033  
Iteration 7: log likelihood = -454.49998  
Iteration 8: log likelihood = -454.4999  
Iteration 9: log likelihood = -454.49989

Multinomial logistic regression	Number of obs	=	497
	LR chi2(38)	=	95.84
	Prob > chi2	=	0.0000
Log likelihood = -454.49989	Pseudo R2	=	0.0954

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9829596	.0162973	-1.04	0.300	.9515308 1.015426
exper		1.051843	.0280991	1.89	0.058	.9981869 1.108384
acctrating2		1.154888	.2501015	0.66	0.506	.7554474 1.76553
pyperstudlep		1.052838	.0558091	0.97	0.331	.9489445 1.168106
pyperstude~s		1.011614	.010608	1.10	0.271	.9910347 1.03262
pyperstudw~e		1.006776	.0502534	0.14	0.892	.9129456 1.11025
pyperstudb~k		.9911205	.0507165	-0.17	0.862	.8965401 1.095679
pyperstudl~o		.9646466	.0679587	-0.51	0.609	.8402368 1.107477
pyperstudm~e		1.013482	.0242777	0.56	0.576	.9669983 1.0622
male		2.328963	.6899251	2.85	0.004	1.303183 4.162172
age		.9612405	.0226793	-1.68	0.094	.9178021 1.006735
Asian		2.27e-07	.0006647	-0.01	0.996	0
Black		2.074952	.912539	1.66	0.097	.8763143 4.913108
Latino		1.06e-06	.0014855	-0.01	0.992	0
Other		4.00e-07	.0011722	-0.01	0.996	0
AsianMale		1.044985	3686.851	0.00	1.000	0
BlackMale		1.074077	.6072729	0.13	0.899	.3546292 3.253094
LatinoMale		1096210	1.53e+09	0.01	0.992	0
OtherMale		1.28e+14	7.83e+17	0.01	0.996	0
3						
tenure		1.003237	.0156522	0.21	0.836	.9730235 1.034389
exper		1.064556	.0277805	2.40	0.017	1.011476 1.120421
acctrating2		1.422993	.2939276	1.71	0.088	.9492531 2.13316
pyperstudlep		.9960486	.0504191	-0.08	0.938	.9019728 1.099937
pyperstude~s		1.045519	.0107038	4.35	0.000	1.024749 1.06671
pyperstudw~e		.9842693	.0461906	-0.34	0.735	.8977761 1.079095
pyperstudb~k		.946065	.0454073	-1.16	0.248	.8611262 1.039382
pyperstudl~o		.9484198	.0621096	-0.81	0.419	.8341758 1.07831



pyperstudm~e		.9939803	.0244333	-0.25	0.806	.9472273	1.043041
male		3.523865	1.027022	4.32	0.000	1.990392	6.238784
age		.9910926	.0226283	-0.39	0.695	.9477197	1.036451
Asian		4.43e-07	.0011862	-0.01	0.996	0	.
Black		2.863888	1.300564	2.32	0.021	1.175991	6.97442
Latino		5.990623	9.802138	1.09	0.274	.2424912	147.9953
Other		2.60e-06	.0069678	-0.00	0.996	0	.
AsianMale		.5844342	1878.854	-0.00	1.000	0	.
BlackMale		.9134433	.5302438	-0.16	0.876	.2927982	2.849671
LatinoMale		.1835317	.3740577	-0.83	0.406	.0033796	9.966745
OtherMale		252103.3	2.48e+09	0.00	0.999	0	.

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -825.59825  
Iteration 1: log likelihood = -765.75007  
Iteration 2: log likelihood = -763.99878  
Iteration 3: log likelihood = -763.89609  
Iteration 4: log likelihood = -763.88501  
Iteration 5: log likelihood = -763.88336  
Iteration 6: log likelihood = -763.88306  
Iteration 7: log likelihood = -763.88299  
Iteration 8: log likelihood = -763.88297

Multinomial logistic regression	Number of obs	=	771
	LR chi2(36)	=	123.43
	Prob > chi2	=	0.0000
Log likelihood = -763.88297	Pseudo R2	=	0.0748

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9906741	.0135235	-0.69	0.492	.9645199 1.017537
exper		1.010842	.0183449	0.59	0.552	.9755189 1.047445
acctrating2		1.204538	.180249	1.24	0.214	.8983495 1.615086
pyperstudlep		1.018299	.0191052	0.97	0.334	.9815336 1.056441
pyperstude~s		.9804001	.0076399	-2.54	0.011	.9655399 .995489
pyperstudw~e		1.018455	.0426308	0.44	0.662	.9382357 1.105533
pyperstudb~k		1.028933	.045325	0.65	0.517	.9438244 1.121716
pyperstudl~o		1.027963	.0434523	0.65	0.514	.9462304 1.116755
pyperstudm~e		1.012571	.0154535	0.82	0.413	.9827308 1.043316
male		1.244642	.2435655	1.12	0.263	.8481487 1.82649
age		.9708539	.0147254	-1.95	0.051	.9424174 1.000148
Asian		8.79e-07	.0013273	-0.01	0.993	0 .
Black		.4500482	.3043116	-1.18	0.238	.1195907 1.693638
Latino		2.488286	1.515392	1.50	0.134	.7542404 8.209014
Other		1.482198	5080.739	0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		8.399356	8.130901	2.20	0.028	1.259651 56.00695
LatinoMale		.2763495	.2435254	-1.46	0.144	.0491316 1.554378
OtherMale		542947.8	1.91e+09	0.00	0.997	0 .
3						
tenure		.9632581	.0128991	-2.80	0.005	.9383052 .9888746
exper		1.112616	.0233783	5.08	0.000	1.067726 1.159393
acctrating2		1.538751	.2486586	2.67	0.008	1.121027 2.11213
pyperstudlep		1.023371	.0196926	1.20	0.230	.9854931 1.062705
pyperstude~s		1.02846	.008146	3.54	0.000	1.012618 1.044551
pyperstudw~e		1.031835	.0400084	0.81	0.419	.9563253 1.113306
pyperstudb~k		1.03152	.0412753	0.78	0.438	.9537129 1.115675
pyperstudl~o		1.008468	.0394764	0.22	0.829	.9339895 1.088886

pyperstudm~e		.9970681	.0153284	-0.19	0.849	.9674731	1.027568
male		2.045515	.4248462	3.45	0.001	1.36148	3.073224
age		.9420292	.0174876	-3.22	0.001	.9083701	.9769355
Asian		1.24e-06	.0019871	-0.01	0.993	0	.
Black		1.180045	.6218005	0.31	0.753	.4201221	3.314528
Latino		.8788487	.7911956	-0.14	0.886	.1505248	5.131213
Other		1.72e+07	3.71e+10	0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		1.028628	.9829035	0.03	0.976	.1580865	6.693021
LatinoMale		.539324	.6582616	-0.51	0.613	.0493096	5.898854
OtherMale		.0297904	68.35292	-0.00	0.999	0	.

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1022.8508  
Iteration 1: log likelihood = -946.94145  
Iteration 2: log likelihood = -943.70513  
Iteration 3: log likelihood = -943.53041  
Iteration 4: log likelihood = -943.50046  
Iteration 5: log likelihood = -943.49379  
Iteration 6: log likelihood = -943.49219  
Iteration 7: log likelihood = -943.49185  
Iteration 8: log likelihood = -943.49178  
Iteration 9: log likelihood = -943.49176

Multinomial logistic regression

Number of obs = 1005  
LR chi2(36) = 158.72  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0776

Log likelihood = -943.49176

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9874591	.0134739	-0.92	0.355	.9614007 1.014224
exper		.9924252	.0160773	-0.47	0.639	.9614093 1.024442
acctrating2		.7662055	.1406097	-1.45	0.147	.5347337 1.097875
pyperstudlep		.9944059	.0212311	-0.26	0.793	.9536524 1.036901
pyperstude~s		.9811933	.0085648	-2.18	0.030	.9645495 .9981244
pyperstudw~e		.9054171	.108357	-0.83	0.406	.7161104 1.144768
pyperstudb~k		.9091589	.1098397	-0.79	0.431	.7174675 1.152066
pyperstudl~o		.9361107	.1150827	-0.54	0.591	.73567 1.191164
pyperstudm~e		1.000057	.0155051	0.00	0.997	.9701244 1.030912
male		1.157298	.2358475	0.72	0.473	.776207 1.725491
age		.9774039	.0138141	-1.62	0.106	.9507004 1.004858
Asian		6.41e-07	.0010875	-0.01	0.993	0 .
Black		1.265951	.4699048	0.64	0.525	.6115918 2.620426
Latino		.6670826	.5983262	-0.45	0.652	.1150037 3.869435
Other		2.25e-06	.001756	-0.02	0.987	0 .
AsianMale		(omitted)				
BlackMale		5.412121	3.085683	2.96	0.003	1.770357 16.54528
LatinoMale		7.561069	11.26159	1.36	0.174	.4081203 140.0807
OtherMale		.6713476	957.3747	-0.00	1.000	0 .
3						
tenure		.9970961	.0101939	-0.28	0.776	.9773153 1.017277
exper		1.018742	.0133388	1.42	0.156	.9929309 1.045224
acctrating2		1.430034	.188875	2.71	0.007	1.10388 1.852553
pyperstudlep		1.037683	.0188527	2.04	0.042	1.001382 1.075299
pyperstude~s		1.030744	.0068673	4.55	0.000	1.017372 1.044292
pyperstudw~e		.9591765	.095484	-0.42	0.675	.7891566 1.165826
pyperstudb~k		.9417455	.0942275	-0.60	0.549	.7740439 1.145781

pyperstudl~o		.9259001	.0943388	-0.76	0.450	.7582917	1.130556
pyperstudm~e		.995802	.0139683	-0.30	0.764	.9687976	1.023559
male		1.730209	.2902091	3.27	0.001	1.245449	2.403649
age		.9948995	.0117277	-0.43	0.664	.9721772	1.018153
Asian		2.45e-06	.0031186	-0.01	0.992	0	.
Black		1.203092	.3956192	0.56	0.574	.6315297	2.291944
Latino		.2553772	.2881829	-1.21	0.226	.0279663	2.332001
Other		1.453423	1.84849	0.29	0.769	.1201762	17.57785
AsianMale		(omitted)					
BlackMale		2.826676	1.552481	1.89	0.058	.9633294	8.294251
LatinoMale		28.32878	45.64105	2.08	0.038	1.204593	666.2165
OtherMale		7.15e-07	.0006383	-0.02	0.987	0	.

-----> region = 8

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -370.60356  
Iteration 1: log likelihood = -321.72819  
Iteration 2: log likelihood = -319.43205  
Iteration 3: log likelihood = -318.22717  
Iteration 4: log likelihood = -318.1472  
Iteration 5: log likelihood = -318.13582  
Iteration 6: log likelihood = -318.13443  
Iteration 7: log likelihood = -318.13416  
Iteration 8: log likelihood = -318.1341  
Iteration 9: log likelihood = -318.13409

Multinomial logistic regression	Number of obs	=	380
	LR chi2(34)	=	104.94
	Prob > chi2	=	0.0000
Log likelihood = -318.13409	Pseudo R2	=	0.1416

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.052844	.0278097	1.95	0.051	.9997246 1.108785
exper		.9588306	.03187	-1.26	0.206	.8983578 1.023374
acctrating2		1.398195	.4527734	1.04	0.301	.741183 2.637606
pyperstudlep		.9632654	.0447316	-0.81	0.420	.8794644 1.055051
pyperstude~s		.9790538	.0164629	-1.26	0.208	.947313 1.011858
pyperstudw~e		.7989927	.1190705	-1.51	0.132	.5966118 1.070025
pyperstudb~k		.8413415	.1267231	-1.15	0.251	.6262737 1.130266
pyperstudl~o		.8282418	.1252835	-1.25	0.213	.6157447 1.114073
pyperstudm~e		.9615615	.0278154	-1.36	0.175	.9085609 1.017654
male		3.154156	1.215668	2.98	0.003	1.481886 6.713537
age		.9924408	.0293148	-0.26	0.797	.9366164 1.051592
Asian		5.03e+08	5.12e+12	0.00	0.998	0 .
Black		2.207403	1.473719	1.19	0.236	.5964825 8.168941
Latino		3.566812	8175.589	0.00	1.000	0 .
Other		2.969756	4.063004	0.80	0.426	.2033148 43.37831
AsianMale		(omitted)				
BlackMale		.9582326	1.175867	-0.03	0.972	.0864857 10.6169
LatinoMale		(omitted)				
OtherMale		7618394	1.96e+10	0.01	0.995	0 .
3						
tenure		1.022572	.0183479	1.24	0.213	.9872363 1.059174
exper		1.014227	.0247381	0.58	0.562	.9668817 1.06389
acctrating2		1.775332	.3998053	2.55	0.011	1.141796 2.76039
pyperstudlep		.9895786	.0313168	-0.33	0.741	.9300636 1.052902
pyperstude~s		1.040194	.0120226	3.41	0.001	1.016895 1.064027

pyperstudw~e		.782208	.1093795	-1.76	0.079	.5946962	1.028844
pyperstudb~k		.768547	.1071022	-1.89	0.059	.5848571	1.00993
pyperstudl~o		.7665165	.1069999	-1.90	0.057	.5830426	1.007727
pyperstudm~e		.976642	.0220637	-1.05	0.295	.9343413	1.020858
male		3.078193	.8898904	3.89	0.000	1.746697	5.424679
age		1.023206	.0243281	0.96	0.335	.9766173	1.072016
Asian		.2731356	4595.606	-0.00	1.000	0	.
Black		2.392259	1.366067	1.53	0.127	.7811618	7.326142
Latino		7584105	8.52e+09	0.01	0.989	0	.
Other		2.42e-06	.0022099	-0.01	0.989	0	.
AsianMale		(omitted)					
BlackMale		1.499473	1.593593	0.38	0.703	.1867704	12.03841
LatinoMale		(omitted)					
OtherMale		3.74e+12	1.02e+16	0.01	0.992	0	.

-----> region = 9

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -231.55137  
Iteration 1: log likelihood = -201.18969  
Iteration 2: log likelihood = -198.72207  
Iteration 3: log likelihood = -198.43924  
Iteration 4: log likelihood = -198.38158  
Iteration 5: log likelihood = -198.36884  
Iteration 6: log likelihood = -198.36674  
Iteration 7: log likelihood = -198.36624  
Iteration 8: log likelihood = -198.36613  
Iteration 9: log likelihood = -198.3661  
Iteration 10: log likelihood = -198.3661

Multinomial logistic regression	Number of obs	=	219
	LR chi2(34)	=	66.37
	Prob > chi2	=	0.0007
Log likelihood = -198.3661	Pseudo R2	=	0.1433

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]		
-----							
1							
tenure		.9993768	.0236739	-0.03	0.979	.9540376	1.046871
exper		.9455545	.0330463	-1.60	0.109	.8829535	1.012594
acctrating2		.6712689	.1994841	-1.34	0.180	.374922	1.201855
pyperstudlep		1.015618	.0632742	0.25	0.804	.8988751	1.147522
pyperstude~s		.9864022	.0133519	-1.01	0.312	.960577	1.012922
pyperstudw~e		1.044166	.1205264	0.37	0.708	.8327543	1.30925
pyperstudb~k		1.032521	.1332329	0.25	0.804	.8017941	1.329643
pyperstudl~o		1.033229	.1207215	0.28	0.780	.8217563	1.299123
pyperstudm~e		1.017138	.0256022	0.68	0.500	.9681765	1.068576
male		1.040191	.3695033	0.11	0.912	.5184938	2.086807
age		1.062784	.0306133	2.11	0.035	1.004445	1.124511
Asian		1.289929	27251.54	0.00	1.000	0	.
Black		.390262	.5268986	-0.70	0.486	.0276781	5.502705
Latino		6.16e+07	3.72e+11	0.00	0.998	0	.
Other		4.51e-08	.0001162	-0.01	0.995	0	.
AsianMale		(omitted)					
BlackMale		4.06e+07	8.49e+10	0.01	0.993	0	.
LatinoMale		5.75e-16	4.18e-12	-0.00	0.996	0	.
OtherMale		(omitted)					
-----							
2							
tenure		.9741643	.0353907	-0.72	0.471	.9072118	1.046058
exper		.9192515	.0406827	-1.90	0.057	.8428753	1.002548
acctrating2		.5561602	.2101927	-1.55	0.121	.2651554	1.166539
pyperstudlep		.9936211	.073606	-0.09	0.931	.85934	1.148885

pyperstude~s		1.007875	.0189219	0.42	0.676	.9714627	1.045652
pyperstudw~e		.7550035	.106846	-1.99	0.047	.5721229	.9963425
pyperstudb~k		.7947205	.1215774	-1.50	0.133	.5888389	1.072586
pyperstudl~o		.7468807	.1083071	-2.01	0.044	.5621035	.9923987
pyperstudm~e		1.020596	.0264655	0.79	0.432	.9700206	1.073808
male		1.16267	.5551548	0.32	0.752	.4560604	2.964084
age		1.024406	.0339464	0.73	0.467	.9599864	1.093147
Asian		3.76e+08	5.45e+12	0.00	0.999	0	.
Black		.4109403	.4580118	-0.80	0.425	.0462462	3.651589
Latino		.4312429	4539.806	-0.00	1.000	0	.
Other		.2927352	.394661	-0.91	0.362	.02084	4.111983
AsianMale		(omitted)					
BlackMale		1.35e+07	2.82e+10	0.01	0.994	0	.
LatinoMale		1.29e-07	.0015216	-0.00	0.999	0	.
OtherMale		(omitted)					

-----+-----  
3 | (base outcome)  
-----+-----

-----> region = 10

Iteration 0: log likelihood = -3561.5192  
Iteration 1: log likelihood = -3356.5853  
Iteration 2: log likelihood = -3352.52  
Iteration 3: log likelihood = -3352.5089  
Iteration 4: log likelihood = -3352.5089

Multinomial logistic regression

Number of obs = 3331  
LR chi2(38) = 418.02  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0587

Log likelihood = -3352.5089

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.015793	.0065062	2.45	0.014	1.00312 1.028625
exper		1.01288	.0085229	1.52	0.128	.9963121 1.029723
acctrating2		.8587046	.0647138	-2.02	0.043	.7407906 .9953874
pyperstudlep		.9999111	.0050244	-0.02	0.986	.9901119 1.009807
pyperstude~s		.9918399	.0036558	-2.22	0.026	.9847006 .999031
pyperstudw~e		.9908861	.0081049	-1.12	0.263	.9751275 1.006899
pyperstudb~k		.9870218	.0086282	-1.49	0.135	.9702549 1.004078
pyperstudl~o		.994514	.0085126	-0.64	0.520	.9779687 1.011339
pyperstudm~e		1.007549	.0061373	1.23	0.217	.9955922 1.01965
male		1.278901	.1357795	2.32	0.020	1.038642 1.574737
age		.968632	.0068146	-4.53	0.000	.9553673 .9820809
Asian		.4069682	.3217489	-1.14	0.255	.0864173 1.91655
Black		1.636729	.233047	3.46	0.001	1.238163 2.163594
Latino		1.737391	.3394428	2.83	0.005	1.184661 2.54801
Other		.4726369	.2666801	-1.33	0.184	.1564036 1.428264
AsianMale		2.99717	3.089664	1.06	0.287	.3974124 22.6038
BlackMale		1.57003	.3461424	2.05	0.041	1.019166 2.41864
LatinoMale		1.376153	.4240497	1.04	0.300	.7522743 2.517429
OtherMale		4.918548	3.632317	2.16	0.031	1.156724 20.91433
3						
tenure		1.023521	.0064198	3.71	0.000	1.011015 1.036181
exper		1.070628	.0103837	7.04	0.000	1.050469 1.091174
acctrating2		1.354797	.1032306	3.99	0.000	1.166853 1.573015
pyperstudlep		1.017275	.0058433	2.98	0.003	1.005887 1.028792
pyperstude~s		1.017387	.003984	4.40	0.000	1.009608 1.025225
pyperstudw~e		1.015776	.0091758	1.73	0.083	.9979501 1.033921
pyperstudb~k		.999688	.009813	-0.03	0.975	.9806388 1.019107

pyperstudl~o		.9858492	.0093493	-1.50	0.133	.9676943	1.004345
pyperstudm~e		.9922866	.0075035	-1.02	0.306	.9776884	1.007103
male		1.065603	.12289	0.55	0.582	.8500242	1.335856
age		.9654047	.0082491	-4.12	0.000	.9493714	.9817087
Asian		1.515394	.9670841	0.65	0.515	.4338219	5.293461
Black		.9617083	.1621645	-0.23	0.817	.6910552	1.338363
Latino		1.799622	.3777471	2.80	0.005	1.192644	2.715511
Other		.3991627	.2362027	-1.55	0.121	.125157	1.273048
AsianMale		1.12351	1.034998	0.13	0.899	.1846864	6.834695
BlackMale		2.101837	.5434258	2.87	0.004	1.266259	3.488795
LatinoMale		1.821869	.6108167	1.79	0.074	.9443552	3.514786
OtherMale		2.003382	1.862219	0.75	0.455	.323996	12.38762

-----> region = 11

Iteration 0: log likelihood = -2446.2189  
Iteration 1: log likelihood = -2319.6715  
Iteration 2: log likelihood = -2317.023  
Iteration 3: log likelihood = -2317.0164  
Iteration 4: log likelihood = -2317.0164

Multinomial logistic regression	Number of obs	=	2269
	LR chi2(38)	=	258.40
	Prob > chi2	=	0.0000
Log likelihood = -2317.0164	Pseudo R2	=	0.0528

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.001022	.0081607	0.13	0.900	.9851547 1.017145
exper		.9930868	.0097337	-0.71	0.479	.9741913 1.012349
acctrating2		1.077284	.1000982	0.80	0.423	.8979223 1.292473
pyperstudlep		1.000141	.0074462	0.02	0.985	.9856529 1.014842
pyperstude~s		1.001127	.0040363	0.28	0.780	.9932471 1.009069
pyperstudw~e		.9937409	.0125831	-0.50	0.620	.969382 1.018712
pyperstudb~k		.9964213	.0136841	-0.26	0.794	.9699588 1.023606
pyperstudl~o		.9977815	.0131496	-0.17	0.866	.9723388 1.02389
pyperstudm~e		.9912652	.0069722	-1.25	0.212	.9776936 1.005025
male		1.064525	.1261175	0.53	0.598	.8439399 1.342766
age		.9882334	.0081139	-1.44	0.149	.9724578 1.004265
Asian		.3646049	.4090912	-0.90	0.369	.0404356 3.287617
Black		1.442477	.3065683	1.72	0.085	.9510514 2.187832
Latino		1.112392	.3245136	0.37	0.715	.6279719 1.970495
Other		.7739204	.4296378	-0.46	0.644	.2607092 2.297398
AsianMale		4.938211	7.179943	1.10	0.272	.2857327 85.34527
BlackMale		1.898789	.6988233	1.74	0.081	.9230039 3.906157
LatinoMale		1.470118	.6046106	0.94	0.349	.6565708 3.291721
OtherMale		1.215186	.9772464	0.24	0.809	.2512527 5.877257
3						
tenure		1.001722	.0074907	0.23	0.818	.9871478 1.016512
exper		1.053477	.0109043	5.03	0.000	1.032321 1.075068
acctrating2		1.553351	.1487207	4.60	0.000	1.287579 1.873981
pyperstudlep		1.010545	.0084225	1.26	0.208	.9941714 1.027189
pyperstude~s		1.02287	.0044976	5.14	0.000	1.014093 1.031723
pyperstudw~e		1.020613	.0147395	1.41	0.158	.9921297 1.049915
pyperstudb~k		1.011347	.0157902	0.72	0.470	.9808671 1.042773
pyperstudl~o		.9998755	.0152637	-0.01	0.993	.9704023 1.030244
pyperstudm~e		.9883166	.0079755	-1.46	0.145	.9728079 1.004073
male		1.363111	.1701788	2.48	0.013	1.067241 1.741006
age		1.002539	.0094161	0.27	0.787	.9842524 1.021165
Asian		1.236425	1.118509	0.23	0.815	.2099634 7.281013

Black		1.267244	.3227976	0.93	0.352	.7691992	2.087766
Latino		2.014461	.5867483	2.40	0.016	1.138232	3.565223
Other		.2953758	.2346397	-1.54	0.125	.0622587	1.40136
AsianMale		.7316312	1.139042	-0.20	0.841	.0346016	15.46992
BlackMale		2.026235	.8446435	1.69	0.090	.8950829	4.586868
LatinoMale		.4213727	.2034726	-1.79	0.073	.1635453	1.085662
OtherMale		3.127335	3.228545	1.10	0.269	.4134523	23.65503

-----> region = 12

```
Iteration 0: log likelihood = -880.19959
Iteration 1: log likelihood = -807.69528
Iteration 2: log likelihood = -805.244
Iteration 3: log likelihood = -804.93596
Iteration 4: log likelihood = -804.88639
Iteration 5: log likelihood = -804.87491
Iteration 6: log likelihood = -804.87209
Iteration 7: log likelihood = -804.87153
Iteration 8: log likelihood = -804.87141
Iteration 9: log likelihood = -804.87138
Iteration 10: log likelihood = -804.87138
```

```
Multinomial logistic regression      Number of obs   =      818
                                     LR chi2(38)        =     150.66
                                     Prob > chi2         =     0.0000
Log likelihood = -804.87138          Pseudo R2        =     0.0856
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9999511	.0125633	-0.00	0.997	.9756283 1.02488
exper		.9998442	.0167694	-0.01	0.993	.9675111 1.033258
acctrating2		1.073007	.1916095	0.39	0.693	.7561389 1.522663
pyperstudlep		1.008753	.0218669	0.40	0.688	.9667918 1.052534
pyperstude~s		.9862816	.0074635	-1.83	0.068	.9717614 1.001019
pyperstudw~e		.8559904	.049119	-2.71	0.007	.7649352 .9578845
pyperstudb~k		.8782574	.0544964	-2.09	0.036	.777686 .9918349
pyperstudl~o		.88078	.0499248	-2.24	0.025	.7881688 .9842731
pyperstudm~e		.9909087	.0094575	-0.96	0.339	.9725446 1.00962
male		.9430064	.2002714	-0.28	0.782	.6219283 1.429845
age		.9713513	.014744	-1.91	0.055	.9428793 1.000683
Asian		2.48e-07	.0004951	-0.01	0.994	0 .
Black		1.358051	.4582586	0.91	0.364	.7009513 2.631141
Latino		.951498	.5835471	-0.08	0.935	.2860038 3.165511
Other		.1610119	.1904645	-1.54	0.123	.0158472 1.635923
AsianMale		.2580038	835.9733	-0.00	1.000	0 .
BlackMale		.9329681	.5497521	-0.12	0.906	.2939651 2.960996
LatinoMale		8.281494	10.27307	1.70	0.088	.7281255 94.19139
OtherMale		2.64e-06	.004312	-0.01	0.994	0 .
-----						
3						
tenure		.9991575	.0110607	-0.08	0.939	.9777124 1.021073
exper		1.039181	.0168317	2.37	0.018	1.00671 1.0727
acctrating2		2.05141	.3115745	4.73	0.000	1.523243 2.762712
pyperstudlep		1.032718	.0216447	1.54	0.125	.9911548 1.076024
pyperstude~s		1.016917	.0078854	2.16	0.031	1.001579 1.032491
pyperstudw~e		.9225466	.0572827	-1.30	0.194	.8168374 1.041936
pyperstudb~k		.9261929	.0612045	-1.16	0.246	.8136778 1.054267
pyperstudl~o		.9063613	.0568684	-1.57	0.117	.8014821 1.024965
pyperstudm~e		.9751472	.0116217	-2.11	0.035	.9526331 .9981934
male		1.529638	.3041608	2.14	0.033	1.035933 2.258634
age		.9797966	.014602	-1.37	0.171	.9515912 1.008838

Asian		3.030091	4.43144	0.76	0.448	.1724218	53.24992
Black		1.161884	.4616815	0.38	0.706	.533252	2.531586
Latino		1.079183	.7076841	0.12	0.907	.2984778	3.901919
Other		1.93e-07	.0002418	-0.01	0.990	0	.
AsianMale		5.87e-08	.0001424	-0.01	0.995	0	.
BlackMale		1.175984	.7117556	0.27	0.789	.359102	3.8511
LatinoMale		1.460137	2.300835	0.24	0.810	.0665425	32.0397
OtherMale		7121096	8.91e+09	0.01	0.990	0	.

-----> region = 13

Iteration 0: log likelihood = -1506.5266  
Iteration 1: log likelihood = -1423.0188  
Iteration 2: log likelihood = -1420.7099  
Iteration 3: log likelihood = -1420.4531  
Iteration 4: log likelihood = -1420.4116  
Iteration 5: log likelihood = -1420.402  
Iteration 6: log likelihood = -1420.3997  
Iteration 7: log likelihood = -1420.3993  
Iteration 8: log likelihood = -1420.3992  
Iteration 9: log likelihood = -1420.3992

Multinomial logistic regression	Number of obs	=	1405
	LR chi2(38)	=	172.25
	Prob > chi2	=	0.0000
Log likelihood = -1420.3992	Pseudo R2	=	0.0572

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.013172	.0109798	1.21	0.227	.9918787 1.034922
exper		.9750626	.0114724	-2.15	0.032	.9528343 .9978094
acctrating2		1.051834	.1327469	0.40	0.689	.8213364 1.347017
pypersistudlep		1.006559	.0082791	0.79	0.427	.9904621 1.022917
pypersistude~s		.9939002	.0067571	-0.90	0.368	.9807443 1.007232
pypersistudw~e		.9491478	.0190762	-2.60	0.009	.912486 .9872826
pypersistudb~k		.9509657	.0226646	-2.11	0.035	.9075655 .9964413
pypersistudl~o		.9552173	.020327	-2.15	0.031	.9161964 .9959
pypersistudm~e		1.002636	.010266	0.26	0.797	.982716 1.022961
male		1.22582	.1986441	1.26	0.209	.8922561 1.684085
age		.9946358	.0100631	-0.53	0.595	.9751068 1.014556
Asian		2.530686	2.364835	0.99	0.320	.4053414 15.79994
Black		2.367502	.6875198	2.97	0.003	1.339993 4.182908
Latino		1.800838	.4074565	2.60	0.009	1.155799 2.805867
Other		4.213215	3.588419	1.69	0.091	.7936709 22.36592
AsianMale		6473037	1.57e+10	0.01	0.995	0 .
BlackMale		1.32088	.6715568	0.55	0.584	.4876395 3.577899
LatinoMale		.9529954	.3480008	-0.13	0.895	.4658666 1.949486
OtherMale		.3035044	.332927	-1.09	0.277	.0353546 2.60546
-----						
3						
tenure		1.028168	.0105177	2.72	0.007	1.007759 1.048991
exper		1.019564	.0128739	1.53	0.125	.9946417 1.045112
acctrating2		1.384654	.1754734	2.57	0.010	1.080118 1.775055
pypersistudlep		1.010416	.0087271	1.20	0.230	.9934552 1.027666
pypersistude~s		1.026712	.0073058	3.70	0.000	1.012493 1.041132
pypersistudw~e		.9601941	.0214395	-1.82	0.069	.9190796 1.003148
pypersistudb~k		.9487389	.0254606	-1.96	0.050	.9001267 .9999765
pypersistudl~o		.9427504	.0222724	-2.50	0.013	.9000926 .9874299
pypersistudm~e		.9685597	.0143123	-2.16	0.031	.9409104 .9970214
male		.9361265	.1564384	-0.39	0.693	.674665 1.298916
age		1.003689	.0113601	0.33	0.745	.9816689 1.026203



Asian		1.080803	1.352608	0.06	0.950	.0929995	12.56065
Black		.4906395	.205686	-1.70	0.089	.2157355	1.115844
Latino		1.318621	.3217954	1.13	0.257	.8173221	2.127387
Other		1.77954	1.825584	0.56	0.574	.2382768	13.29027
AsianMale		2.13658	8482.362	0.00	1.000	0	.
BlackMale		5.648086	3.592492	2.72	0.006	1.623643	19.64772
LatinoMale		1.312136	.5302809	0.67	0.501	.594262	2.897208
OtherMale		4.27e-07	.0003809	-0.02	0.987	0	.

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -247.82971  
Iteration 1: log likelihood = -219.04273  
Iteration 2: log likelihood = -211.57289  
Iteration 3: log likelihood = -209.21491  
Iteration 4: log likelihood = -208.70513  
Iteration 5: log likelihood = -208.67458  
Iteration 6: log likelihood = -208.66848  
Iteration 7: log likelihood = -208.66714  
Iteration 8: log likelihood = -208.66684  
Iteration 9: log likelihood = -208.66677  
Iteration 10: log likelihood = -208.66676

Multinomial logistic regression	Number of obs	=	265
	LR chi2(32)	=	78.33
	Prob > chi2	=	0.0000
Log likelihood = -208.66676	Pseudo R2	=	0.1580

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.03635	.0370709	1.00	0.318	.9661813 1.111615
exper		.9835681	.0521427	-0.31	0.755	.8865005 1.091264
acctrating2		.4076908	.2161623	-1.69	0.091	.1442162 1.152518
pyperstudlep		.9700185	.1323894	-0.22	0.824	.742347 1.267515
pyperstude~s		.9479177	.0219595	-2.31	0.021	.9058404 .9919495
pyperstudw~e		1.03138	.1026359	0.31	0.756	.8486195 1.253501
pyperstudb~k		1.074916	.132421	0.59	0.558	.8443319 1.368471
pyperstudl~o		1.06653	.1067616	0.64	0.520	.8765286 1.297716
pyperstudm~e		1.056567	.0441707	1.32	0.188	.9734462 1.146786
male		3.03141	1.824076	1.84	0.065	.9320845 9.85903
age		1.015655	.0475882	0.33	0.740	.9265385 1.113343
Asian		(omitted)				
Black		4.61e-06	.0058083	-0.01	0.992	0 .
Latino		6.579485	9.866084	1.26	0.209	.348174 124.3333
Other		18.52977	53896.15	0.00	0.999	0 .
AsianMale		(omitted)				
BlackMale		121092.5	2.91e+08	0.00	0.996	0 .
LatinoMale		1.103598	1.859015	0.06	0.953	.0406404 29.96837
OtherMale		(omitted)				
3						
tenure		1.005978	.020742	0.29	0.773	.9661348 1.047464
exper		1.071209	.0332691	2.21	0.027	1.007948 1.138441
acctrating2		1.529025	.3378159	1.92	0.055	.9916402 2.357628
pyperstudlep		1.052236	.0632128	0.85	0.397	.9353567 1.183719
pyperstude~s		1.017092	.0123547	1.40	0.163	.993163 1.041597
pyperstudw~e		1.258191	.2344649	1.23	0.218	.8732217 1.812878
pyperstudb~k		1.218319	.2398145	1.00	0.316	.8283429 1.791892

pyperstudl~o		1.261832	.2354535	1.25	0.213	.8753272	1.819001
pyperstudm~e		1.013422	.0242588	0.56	0.578	.9669737	1.062101
male		2.129628	.6786585	2.37	0.018	1.140374	3.977042
age		.9514641	.0253552	-1.87	0.062	.9030443	1.00248
Asian		(omitted)					
Black		3.151115	4.687043	0.77	0.440	.1707524	58.15158
Latino		.9877025	1.121635	-0.01	0.991	.1066601	9.146407
Other		4164533	4.82e+09	0.01	0.989	0	.
AsianMale		(omitted)					
BlackMale		286754.3	2.34e+08	0.02	0.988	0	.
LatinoMale		.2945075	.4985809	-0.72	0.470	.0106677	8.130586
OtherMale		(omitted)					

-----> region = 15

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -283.69024  
Iteration 1: log likelihood = -251.74922  
Iteration 2: log likelihood = -248.47897  
Iteration 3: log likelihood = -248.31378  
Iteration 4: log likelihood = -248.2885  
Iteration 5: log likelihood = -248.28392  
Iteration 6: log likelihood = -248.28291  
Iteration 7: log likelihood = -248.28266  
Iteration 8: log likelihood = -248.28261  
Iteration 9: log likelihood = -248.2826

Multinomial logistic regression	Number of obs	=	280
	LR chi2(30)	=	70.82
	Prob > chi2	=	0.0000
Log likelihood = -248.2826	Pseudo R2	=	0.1248

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.069691	.0412693	1.75	0.081	.9917869 1.153714
exper		.9254502	.0409503	-1.75	0.080	.8485709 1.009295
acctrating2		.597705	.2720128	-1.13	0.258	.2449678 1.45836
pyperstudlep		.9638911	.0391769	-0.90	0.366	.8900846 1.043818
pyperstude~s		1.005519	.0166634	0.33	0.740	.973384 1.038715
pyperstudw~e		.554568	.1718415	-1.90	0.057	.3021321 1.017918
pyperstudb~k		.552299	.1827504	-1.79	0.073	.2887486 1.056401
pyperstudl~o		.5543437	.1724809	-1.90	0.058	.3012538 1.02006
pyperstudm~e		1.010448	.0256346	0.41	0.682	.961434 1.061961
male		1.880205	1.02154	1.16	0.245	.6482405 5.453487
age		1.036046	.0327604	1.12	0.263	.9737863 1.102287
Asian		(omitted)				
Black		2.71e+08	7.93e+11	0.01	0.995	0
Latino		5.607428	3.779204	2.56	0.011	1.496527 21.01081
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		.0093126	27.85033	-0.00	0.999	0
LatinoMale		.5439562	.4769744	-0.69	0.487	.0975389 3.033543
OtherMale		(omitted)				
3						
tenure		.9732322	.0194892	-1.35	0.175	.9357741 1.01219
exper		1.077985	.030845	2.62	0.009	1.019194 1.140168
acctrating2		1.492711	.3820883	1.57	0.118	.9038455 2.465229

pyperstudlep		1.025934	.0287639	0.91	0.361	.971079	1.083888
pyperstude~s		1.014386	.0116469	1.24	0.213	.9918139	1.037473
pyperstudw~e		.857561	.1957862	-0.67	0.501	.5481895	1.341527
pyperstudb~k		.827792	.2012101	-0.78	0.437	.5140685	1.332973
pyperstudl~o		.8447319	.1940089	-0.73	0.463	.5385477	1.324993
pyperstudm~e		1.020755	.0174107	1.20	0.228	.9871949	1.055456
male		.9794423	.3233326	-0.06	0.950	.5128389	1.870582
age		.9759619	.0247674	-0.96	0.338	.9286062	1.025733
Asian		(omitted)					
Black		.9645918	3785.711	-0.00	1.000	0	.
Latino		.7194896	.4173189	-0.57	0.570	.2308394	2.242534
Other		(omitted)					
AsianMale		(omitted)					
BlackMale		1672943	6.65e+09	0.00	0.997	0	.
LatinoMale		1.701531	1.381295	0.65	0.513	.3466077	8.352981
OtherMale		(omitted)					

-----> region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -496.28686  
Iteration 1: log likelihood = -434.9081  
Iteration 2: log likelihood = -432.68307  
Iteration 3: log likelihood = -432.51626  
Iteration 4: log likelihood = -432.48134  
Iteration 5: log likelihood = -432.47371  
Iteration 6: log likelihood = -432.47232  
Iteration 7: log likelihood = -432.47218  
Iteration 8: log likelihood = -432.47215  
Iteration 9: log likelihood = -432.47214

Multinomial logistic regression	Number of obs	=	461
	LR chi2(34)	=	127.63
	Prob > chi2	=	0.0000
Log likelihood = -432.47214	Pseudo R2	=	0.1286

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.037933	.0230528	1.68	0.094	.9937197 1.084114
exper		.9248569	.0261418	-2.76	0.006	.8750134 .9775396
acctrating2		.6809814	.1583404	-1.65	0.098	.4317333 1.074125
pyperstudlep		1.005179	.0205858	0.25	0.801	.965631 1.046348
pyperstude~s		.978226	.0091934	-2.34	0.019	.9603722 .9964117
pyperstudw~e		.9548322	.0465939	-0.95	0.344	.867741 1.050664
pyperstudb~k		.984616	.0525772	-0.29	0.772	.8867757 1.093251
pyperstudl~o		.9693222	.046613	-0.65	0.517	.8821357 1.065126
pyperstudm~e		1.054202	.020381	2.73	0.006	1.015003 1.094915
male		1.065634	.2979809	0.23	0.820	.616011 1.843434
age		.9871821	.0218188	-0.58	0.559	.9453312 1.030886
Asian		(omitted)				
Black		.4381034	.5295071	-0.68	0.495	.0410001 4.681323
Latino		3.950138	2.708459	2.00	0.045	1.030329 15.14429
Other		.5802064	.7768222	-0.41	0.684	.0420661 8.002628
AsianMale		(omitted)				
BlackMale		3.93e-07	.0004885	-0.01	0.991	0 .
LatinoMale		.2214436	.2575026	-1.30	0.195	.0226702 2.163073
OtherMale		2.70e+08	1.24e+12	0.00	0.997	0 .
3						
tenure		1.015328	.0168504	0.92	0.359	.9828336 1.048898

exper		1.047184	.0257077	1.88	0.060	.9979907	1.098802
acctrating2		1.715606	.3201945	2.89	0.004	1.190014	2.473334
pyperstudlep		1.055966	.0187359	3.07	0.002	1.019875	1.093334
pyperstude~s		1.007621	.0077898	0.98	0.326	.9924688	1.023005
pyperstudw~e		1.02106	.0480595	0.44	0.658	.931079	1.119736
pyperstudb~k		1.022391	.0515579	0.44	0.661	.9261723	1.128605
pyperstudl~o		1.006037	.0465565	0.13	0.897	.9188036	1.101552
pyperstudm~e		1.00361	.0198735	0.18	0.856	.9654052	1.043328
male		2.44049	.6162621	3.53	0.000	1.487768	4.003306
age		.9572852	.0211411	-1.98	0.048	.9167334	.9996308
Asian		(omitted)					
Black		1.476934	1.471286	0.39	0.695	.2096111	10.40658
Latino		1.19001	.9375842	0.22	0.825	.2540427	5.574355
Other		4.87e-07	.0005263	-0.01	0.989	0	.
AsianMale		(omitted)					
BlackMale		.214017	.3617747	-0.91	0.362	.0077907	5.879237
LatinoMale		1.212845	1.364105	0.17	0.864	.1337969	10.99422
OtherMale		854776.8	5.94e+09	0.00	0.998	0	.

-----> region = 17

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -457.29442  
Iteration 1: log likelihood = -408.24065  
Iteration 2: log likelihood = -406.42032  
Iteration 3: log likelihood = -406.33105  
Iteration 4: log likelihood = -406.31153  
Iteration 5: log likelihood = -406.30676  
Iteration 6: log likelihood = -406.30583  
Iteration 7: log likelihood = -406.30565  
Iteration 8: log likelihood = -406.30564

Multinomial logistic regression	Number of obs	=	450
	LR chi2(32)	=	101.98
	Prob > chi2	=	0.0000
Log likelihood = -406.30564	Pseudo R2	=	0.1115

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9931666	.0221382	-0.31	0.758	.9507107 1.037519
exper		1.034447	.0295723	1.18	0.236	.9780802 1.094062
acctrating2		1.103939	.3359646	0.32	0.745	.6079901 2.004441
pyperstudlep		.9372971	.0470592	-1.29	0.197	.8494556 1.034222
pyperstude~s		.9577112	.0130942	-3.16	0.002	.9323879 .9837223
pyperstudw~e		1.085428	.1477889	0.60	0.547	.8311969 1.417419
pyperstudb~k		1.125618	.1544847	0.86	0.389	.8601389 1.473037
pyperstudl~o		1.127514	.1548007	0.87	0.382	.8615041 1.475661
pyperstudm~e		1.008229	.0256656	0.32	0.747	.9591602 1.059809
male		1.584274	.5063637	1.44	0.150	.8467798 2.964083
age		.979476	.0251357	-0.81	0.419	.9314294 1.030001
Asian		9.52e-07	.0017393	-0.01	0.994	0 .
Black		2.33e-06	.0016151	-0.02	0.985	0 .
Latino		.4824782	.5279839	-0.67	0.505	.0564929 4.120613
Other		(omitted)				
AsianMale		(omitted)				
BlackMale		697335.5	4.84e+08	0.02	0.985	0 .
LatinoMale		4.698084	6.007392	1.21	0.226	.3832639 57.58953
OtherMale		(omitted)				

```

3      |
   tenure | .9916279 .0162173 -0.51 0.607 .9603465 1.023928
   exper | 1.058434 .0232137 2.59 0.010 1.0139 1.104924
   acctrating2 | 1.226814 .2687129 0.93 0.351 .7986145 1.884604
pyperstudlep | .9973945 .021557 -0.12 0.904 .956026 1.040553
pyperstude~s | 1.023456 .0107785 2.20 0.028 1.002547 1.044801
pyperstudw~e | 1.05924 .1182773 0.52 0.606 .8510342 1.318383
pyperstudb~k | 1.040914 .1178674 0.35 0.723 .8337376 1.299572
pyperstudl~o | 1.06116 .1191749 0.53 0.597 .8515023 1.322441
pyperstudm~e | .9520486 .0189447 -2.47 0.014 .9156324 .9899131
   male | 2.849515 .7319793 4.08 0.000 1.722327 4.714399
   age | .9852048 .0198858 -0.74 0.460 .9469902 1.024961
   Asian | 4.73e-06 .0056813 -0.01 0.992 0 .
   Black | 1.933611 2.466266 0.52 0.605 .1587398 23.55334
   Latino | 2.749924 1.526927 1.82 0.068 .9261496 8.165073
   Other | (omitted)
   AsianMale | (omitted)
   BlackMale | 2.326662 4.03505 0.49 0.626 .0777183 69.65353
   LatinoMale | .986081 .7896127 -0.02 0.986 .2052607 4.737175
   OtherMale | (omitted)
-----

```

```

-----> region = 18

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -410.74157
Iteration 1: log likelihood = -374.41662
Iteration 2: log likelihood = -373.19796
Iteration 3: log likelihood = -373.17397
Iteration 4: log likelihood = -373.16862
Iteration 5: log likelihood = -373.16749
Iteration 6: log likelihood = -373.16729
Iteration 7: log likelihood = -373.16726

```

```

Multinomial logistic regression      Number of obs   =      394
                                      LR chi2(34)        =      75.15
                                      Prob > chi2         =      0.0001
Log likelihood = -373.16726          Pseudo R2        =      0.0915

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          |      (base outcome)
-----+-----
2          |
   tenure | .981199 .0226225 -0.82 0.410 .9378465 1.026555
   exper | 1.005453 .0269439 0.20 0.839 .9540066 1.059673
   acctrating2 | .5390733 .180107 -1.85 0.064 .2800639 1.03762
pyperstudlep | 1.043892 .0200843 2.23 0.026 1.00526 1.084008
pyperstude~s | .9763804 .0145602 -1.60 0.109 .948256 1.005339
pyperstudw~e | .6544253 .1264682 -2.19 0.028 .4480895 .9557745
pyperstudb~k | .7030872 .1372259 -1.80 0.071 .4795953 1.030726
pyperstudl~o | .664854 .1274266 -2.13 0.033 .4566507 .9679844
pyperstudm~e | 1.028776 .0172477 1.69 0.091 .9955202 1.063142
   male | .9561106 .3379417 -0.13 0.899 .4782412 1.911478
   age | .9882548 .0221267 -0.53 0.598 .9458251 1.032588
   Asian | (omitted)
   Black | 6.305898 7.509864 1.55 0.122 .6109815 65.08274
   Latino | 1.868461 .9146196 1.28 0.202 .715842 4.876978
   Other | 1.115944 1437.578 0.00 1.000 0 .
   AsianMale | (omitted)
   BlackMale | .7342217 1.284922 -0.18 0.860 .0237785 22.67092
   LatinoMale | 1.304438 .8500792 0.41 0.683 .3636723 4.678821
   OtherMale | 9390699 1.81e+10 0.01 0.993 0 .
-----

```

```

3      |
      tenure | 1.006157 .0181082 0.34 0.733 .9712843 1.042282
      exper | 1.048535 .0255283 1.95 0.052 .9996752 1.099782
      acctrating2 | 1.969491 .4937685 2.70 0.007 1.204898 3.219272
      pyperstudlep | 1.017575 .0178458 0.99 0.320 .9831924 1.05316
      pyperstude~s | 1.019654 .0122659 1.62 0.106 .9958947 1.043981
      pyperstudw~e | .9520687 .1717231 -0.27 0.785 .668557 1.355808
      pyperstudb~k | .9629889 .1778388 -0.20 0.838 .6705431 1.38298
      pyperstudl~o | .9435586 .1695017 -0.32 0.746 .6635265 1.341774
      pyperstudm~e | 1.000299 .0176944 0.02 0.987 .9662125 1.035587
      male | 1.540749 .4505235 1.48 0.139 .8686317 2.732926
      age | .964062 .0212458 -1.66 0.097 .9233074 1.006616
      Asian | (omitted)
      Black | 5.494867 6.578324 1.42 0.155 .5259191 57.41105
      Latino | 1.22304 .5349447 0.46 0.645 .5189602 2.882353
      Other | 344382.9 2.43e+08 0.02 0.986 0 .
      AsianMale | (omitted)
      BlackMale | .4560498 .7941998 -0.45 0.652 .0150198 13.84712
      LatinoMale | .9611483 .5550222 -0.07 0.945 .3099258 2.980733
      OtherMale | 3.42e-06 .0081193 -0.01 0.996 0 .
-----

```

```

-----> region = 19

```

```

Iteration 0: log likelihood = -797.61835
Iteration 1: log likelihood = -728.84813
Iteration 2: log likelihood = -725.6924
Iteration 3: log likelihood = -725.5741
Iteration 4: log likelihood = -725.54653
Iteration 5: log likelihood = -725.54066
Iteration 6: log likelihood = -725.53943
Iteration 7: log likelihood = -725.53916
Iteration 8: log likelihood = -725.53909
Iteration 9: log likelihood = -725.53908

```

```

Multinomial logistic regression      Number of obs   =      736
                                     LR chi2(38)        =     144.16
                                     Prob > chi2         =     0.0000
Log likelihood = -725.53908          Pseudo R2        =     0.0904

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          |      (base outcome)
-----+-----
2      tenure | 1.011387 .0139459 0.82 0.412 .9844198 1.039094
      exper | 1.02535 .0179784 1.43 0.153 .9907116 1.061199
      acctrating2 | .8075625 .1673089 -1.03 0.302 .5380544 1.212065
      pyperstudlep | 1.007777 .0058341 1.34 0.181 .9964067 1.019277
      pyperstude~s | .9969452 .0097222 -0.31 0.754 .978071 1.016184
      pyperstudw~e | .9676101 .1037279 -0.31 0.759 .7842446 1.193849
      pyperstudb~k | 1.024192 .1191688 0.21 0.837 .8153434 1.286536
      pyperstudl~o | .9777214 .1008454 -0.22 0.827 .7987656 1.19677
      pyperstudm~e | 1.017774 .0127031 1.41 0.158 .9931787 1.042979
      male | 1.193215 .3567324 0.59 0.555 .6641059 2.143879
      age | .9814145 .0142976 -1.29 0.198 .953788 1.009841
      Asian | 3508515 6.18e+09 0.01 0.993 0 .
      Black | .1637242 .1773293 -1.67 0.095 .0195965 1.367877
      Latino | 1.308059 .3319581 1.06 0.290 .7954451 2.15102
      Other | 1.415092 2.061405 0.24 0.812 .0814359 24.58974
      AsianMale | 1.33e-07 .0005681 -0.00 0.997 0 .
      BlackMale | 4.658763 5.968901 1.20 0.230 .3781762 57.39144
      LatinoMale | .8989863 .3361947 -0.28 0.776 .431945 1.871017
      OtherMale | .8686317 1.797194 -0.07 0.946 .0150558 50.11484
-----

```

```

3      |
      tenure |      1.02684      .0150144      1.81      0.070      .9978302      1.056694
      exper |      1.107653      .0239829      4.72      0.000      1.06163      1.15567
      acctrating2 |      1.094763      .2466345      0.40      0.688      .7039743      1.702487
      pyperstudlep |      1.034141      .0070792      4.90      0.000      1.020359      1.04811
      pyperstude~s |      1.015098      .0112628      1.35      0.177      .9932612      1.037414
      pyperstudw~e |      1.029077      .1288653      0.23      0.819      .8051135      1.315343
      pyperstudb~k |      1.055859      .1451649      0.40      0.693      .8064522      1.3824
      pyperstudl~o |      .992283      .1205493      -0.06      0.949      .7820342      1.259057
      pyperstudm~e |      .9794268      .018069      -1.13      0.260      .9446448      1.015489
      male |      1.24472      .4176062      0.65      0.514      .644899      2.402434
      age |      .9692202      .0184099      -1.65      0.100      .9338008      1.005983
      Asian |      1.233118      3446.58      0.00      1.000      0      .
      Black |      .5971292      .4524748      -0.68      0.496      .1352291      2.636735
      Latino |      .9475704      .2722365      -0.19      0.851      .5395873      1.66403
      Other |      1.48e-06      .0014112      -0.01      0.989      0      .
      AsianMale |      553275.1      2.15e+09      0.00      0.997      0      .
      BlackMale |      3.08656      3.208479      1.08      0.278      .4023933      23.67547
      LatinoMale |      .8227864      .3576121      -0.45      0.654      .3510143      1.928632
      OtherMale |      2.621367      3531.519      0.00      0.999      0      .
-----

```

```

-----> region = 20

```

```

Iteration 0: log likelihood = -2077.1174
Iteration 1: log likelihood = -1971.4581
Iteration 2: log likelihood = -1967.0072
Iteration 3: log likelihood = -1966.5191
Iteration 4: log likelihood = -1966.4382
Iteration 5: log likelihood = -1966.4299
Iteration 6: log likelihood = -1966.4282
Iteration 7: log likelihood = -1966.4278
Iteration 8: log likelihood = -1966.4277
Iteration 9: log likelihood = -1966.4277

```

```

Multinomial logistic regression      Number of obs      =      2049
                                      LR chi2(38)          =      221.38
                                      Prob > chi2          =      0.0000
Log likelihood = -1966.4277          Pseudo R2          =      0.0533

```

```

-----
      admin |      RRR      Std. Err.      z      P>|z|      [95% Conf. Interval]
-----+-----
1      |      (base outcome)
-----+-----
2      |
      tenure |      1.004926      .0082735      0.60      0.551      .9888408      1.021274
      exper |      1.010247      .0106496      0.97      0.333      .9895882      1.031337
      acctrating2 |      1.009711      .1105063      0.09      0.930      .8147765      1.251283
      pyperstudlep |      .9843078      .0076317      -2.04      0.041      .969463      .99938
      pyperstude~s |      .9961943      .0058443      -0.65      0.516      .9848052      1.007715
      pyperstudw~e |      1.024491      .0437055      0.57      0.571      .9423132      1.113835
      pyperstudb~k |      1.022835      .0463411      0.50      0.618      .9359242      1.117817
      pyperstudl~o |      1.029656      .0446582      0.67      0.500      .9457444      1.121012
      pyperstudm~e |      1.005948      .0075232      0.79      0.428      .9913103      1.020802
      male |      1.680217      .2494123      3.50      0.000      1.256066      2.247595
      age |      .971411      .0087693      -3.21      0.001      .9543747      .9887515
      Asian |      .3630263      .3954288      -0.93      0.352      .04293      3.069835
      Black |      1.497138      .4354046      1.39      0.165      .8466644      2.647355
      Latino |      1.219061      .1940307      1.24      0.213      .89237      1.665353
      Other |      1.363071      1.006957      0.42      0.675      .3204058      5.798777
      AsianMale |      3.51e+07      1.05e+11      0.01      0.995      0      .
      BlackMale |      1.048934      .4747501      0.11      0.916      .4320081      2.546854
      LatinoMale |      1.019835      .241206      0.08      0.934      .6415169      1.621256
      OtherMale |      3.66e-07      .0002839      -0.02      0.985      0      .
-----

```

```

3      |
      tenure |      1.01631      .0083066      1.98      0.048      1.000159      1.032722
      exper |      1.088385      .0144757      6.37      0.000      1.06038      1.11713
      acctrating2 |      1.157832      .1375379      1.23      0.217      .9173436      1.461366
      pyperstudlep |      1.016348      .0072409      2.28      0.023      1.002255      1.03064
      pyperstude~s |      1.032712      .0065863      5.05      0.000      1.019883      1.045702
      pyperstudw~e |      1.006626      .0450925      0.15      0.883      .9220145      1.099001
      pyperstudb~k |      .9743111      .0464333      -0.55      0.585      .8874246      1.069705
      pyperstudl~o |      .9751667      .0445102      -0.55      0.582      .8917167      1.066426
      pyperstudm~e |      .9834813      .0096246      -1.70      0.089      .9647972      1.002527
      male |      1.50813      .2490793      2.49      0.013      1.09108      2.084593
      age |      .948463      .0115362      -4.35      0.000      .9261198      .9713453
      Asian |      6.22e-07      .0005435      -0.02      0.987      0      .
      Black |      1.482746      .4602828      1.27      0.204      .8069221      2.724594
      Latino |      .8986566      .1545878      -0.62      0.534      .6414595      1.258978
      Other |      1.03e-06      .0008251      -0.02      0.986      0      .
      AsianMale |      1388560      7.92e+09      0.00      0.998      0      .
      BlackMale |      .711806      .3997958      -0.61      0.545      .2367404      2.140183
      LatinoMale |      1.004027      .2696508      0.01      0.988      .5931148      1.699621
      OtherMale |      1627215      1.31e+09      0.02      0.986      0      .
-----
by pyschooltype:
. bsort pyschooltype: mlogit admin tenure exper acctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlati
> no pyperstudmobile male age Asian Black Latino Other AsianMale BlackMale LatinoMale
OtherMale if validcert2004a==1, rrr iter(20)

```

```

-----> pyschooltype =

```

```

note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -296.6778
Iteration 1: log likelihood = -261.13994
Iteration 2: log likelihood = -259.6833
Iteration 3: log likelihood = -259.45168
Iteration 4: log likelihood = -259.40644
Iteration 5: log likelihood = -259.39654
Iteration 6: log likelihood = -259.39493
Iteration 7: log likelihood = -259.39454
Iteration 8: log likelihood = -259.39446
Iteration 9: log likelihood = -259.39444

```

```

Multinomial logistic regression      Number of obs   =      273
                                      LR chi2(36)       =      74.57
                                      Prob > chi2       =      0.0002
Log likelihood = -259.39444          Pseudo R2       =      0.1257

```

```

-----
      admin |      RRR      Std. Err.      z      P>|z|      [95% Conf. Interval]
-----+-----
1      |      (base outcome)
-----+-----
2      |
      tenure |      .9968326      .0258791      -0.12      0.903      .9473794      1.048867
      exper |      1.110009      .035792      3.24      0.001      1.042029      1.182424
      acctrating2 |      .886691      .2197995      -0.49      0.628      .5454691      1.441367
      pyperstudlep |      1.011792      .0123952      0.96      0.339      .9877876      1.036381
      pyperstude~s |      1.012302      .0145517      0.85      0.395      .9841788      1.041228
      pyperstudw~e |      .9524579      .0462793      -1.00      0.316      .8659374      1.047623
      pyperstudb~k |      .9615611      .0520048      -0.72      0.469      .8648498      1.069087
      pyperstudl~o |      .9398155      .0481558      -1.21      0.226      .8500166      1.039101
      pyperstudm~e |      1.003279      .0165372      0.20      0.843      .9713844      1.03622
      male |      .8554264      .3661494      -0.36      0.715      .3696933      1.979355
      age |      .9464729      .0278632      -1.87      0.062      .8934077      1.00269
      Asian |      1.25e-06      .0012638      -0.01      0.989      0      .
      Black |      1.10794      .8665692      0.13      0.896      .2391961      5.131907
      Latino |      2.398221      1.16079      1.81      0.071      .9287294      6.192828

```



Other		2.486271	3.638125	0.62	0.534	.1412529	43.76223
AsianMale		812531.6	3.20e+09	0.00	0.997	0	.
BlackMale		976303.6	5.53e+08	0.02	0.981	0	.
LatinoMale		.8356123	.7049914	-0.21	0.831	.1599046	4.366652
OtherMale		(omitted)					
-----							
3							
tenure		1.055693	.0278292	2.06	0.040	1.002534	1.111671
exper		1.085347	.0366101	2.43	0.015	1.015913	1.159526
acctrating2		.729985	.1969752	-1.17	0.243	.4301603	1.23879
pyperstudlep		1.02684	.0141065	1.93	0.054	.9995603	1.054864
pyperstude~s		1.018027	.0150657	1.21	0.227	.9889232	1.047988
pyperstudw~e		1.051415	.061185	0.86	0.389	.9380802	1.178441
pyperstudb~k		1.035655	.066798	0.54	0.587	.9126709	1.175212
pyperstudl~o		1.014678	.0619314	0.24	0.811	.9002737	1.14362
pyperstudm~e		.9995409	.0172427	-0.03	0.979	.9663107	1.033914
male		.7657492	.3604793	-0.57	0.571	.3043549	1.926606
age		.9929364	.0298483	-0.24	0.814	.9361249	1.053196
Asian		6.493524	10.18187	1.19	0.233	.3004649	140.3354
Black		2.006892	1.650115	0.85	0.397	.4005395	10.05547
Latino		1.316409	.7545129	0.48	0.631	.4280716	4.048233
Other		1.670917	2.579169	0.33	0.739	.0811092	34.42228
AsianMale		1188086	2.95e+09	0.01	0.996	0	.
BlackMale		1372250	7.77e+08	0.02	0.980	0	.
LatinoMale		2.920701	2.648589	1.18	0.237	.493843	17.27369
OtherMale		(omitted)					
-----							

-----> pyschooltype = B

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -347.28728  
Iteration 1: log likelihood = -320.30661  
Iteration 2: log likelihood = -308.40708  
Iteration 3: log likelihood = -307.00325  
Iteration 4: log likelihood = -306.97715  
Iteration 5: log likelihood = -306.97334  
Iteration 6: log likelihood = -306.97273  
Iteration 7: log likelihood = -306.97258  
Iteration 8: log likelihood = -306.97255  
Iteration 9: log likelihood = -306.97254

Multinomial logistic regression	Number of obs	=	400
	LR chi2(34)	=	80.63
	Prob > chi2	=	0.0000
Log likelihood = -306.97254	Pseudo R2	=	0.1161

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
1	(base outcome)					
-----						
2						
tenure	1.083523	.0388273	2.24	0.025	1.010033	1.162359
exper	.9398735	.0374048	-1.56	0.119	.8693479	1.016121
acctrating2	.6561453	.2773947	-1.00	0.319	.2865118	1.502649
pyperstudlep	.9803703	.0328218	-0.59	0.554	.9181058	1.046857
pyperstude~s	.9521218	.0157297	-2.97	0.003	.921786	.9834559
pyperstudw~e	.9668746	.0323943	-1.01	0.315	.9054226	1.032497
pyperstudb~k	1.04014	.0438985	0.93	0.351	.9575631	1.129838
pyperstudl~o	1.012597	.0400677	0.32	0.752	.9370338	1.094254
pyperstudm~e	.9697719	.0177542	-1.68	0.094	.9355912	1.005201
male	.5787948	.2752106	-1.15	0.250	.2279227	1.469812
age	.9651904	.0290329	-1.18	0.239	.909932	1.023805
Asian	3.34e-06	.0052531	-0.01	0.994	0	.

Black		.4854336	.4123694	-0.85	0.395	.0918429	2.56575
Latino		1.511939	1.611019	0.39	0.698	.187306	12.20441
Other		3.655832	7.889775	0.60	0.548	.0532083	251.1846
AsianMale		(omitted)					
BlackMale		2.23866	2.796874	0.65	0.519	.1934361	25.90828
LatinoMale		1.339753	2.194656	0.18	0.858	.0540341	33.21863
OtherMale		(omitted)					
-----							
3							
tenure		.9751453	.0168405	-1.46	0.145	.9426908	1.008717
exper		1.015603	.018843	0.83	0.404	.9793354	1.053215
acctrating2		.7953362	.1724195	-1.06	0.291	.5200203	1.216414
pyperstudlep		1.031775	.0177425	1.82	0.069	.9975799	1.067142
pyperstude~s		.9877794	.008389	-1.45	0.148	.9714735	1.004359
pyperstudw~e		1.049137	.0430781	1.17	0.243	.9680137	1.137059
pyperstudb~k		1.069458	.0467356	1.54	0.124	.9816707	1.165095
pyperstudl~o		1.054984	.0446884	1.26	0.206	.9709337	1.14631
pyperstudm~e		.9834401	.011398	-1.44	0.150	.9613523	1.006035
male		1.171958	.2907666	0.64	0.522	.720653	1.905891
age		.9728294	.0165399	-1.62	0.105	.940946	1.005793
Asian		4.07e-06	.0036037	-0.01	0.989	0	.
Black		.3986581	.3526861	-1.04	0.299	.0703974	2.257586
Latino		.2202667	.2609639	-1.28	0.202	.0216011	2.24606
Other		1.389374	2.023363	0.23	0.821	.080021	24.12317
AsianMale		(omitted)					
BlackMale		1.723092	2.066115	0.45	0.650	.1643063	18.07018
LatinoMale		2.367131	3.457919	0.59	0.555	.1351395	41.46315
OtherMale		(omitted)					
-----							

-----> pyschooltype = E

Iteration 0: log likelihood = -10521.777  
Iteration 1: log likelihood = -9932.6355  
Iteration 2: log likelihood = -9927.1161  
Iteration 3: log likelihood = -9927.1116  
Iteration 4: log likelihood = -9927.1116

Multinomial logistic regression	Number of obs	=	9745
	LR chi2(38)	=	1189.33
	Prob > chi2	=	0.0000
Log likelihood = -9927.1116	Pseudo R2	=	0.0565

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.010423	.0043361	2.42	0.016	1.00196 1.018957
exper		1.007163	.0055972	1.28	0.199	.9962521 1.018193
acctrating2		.9096854	.0401574	-2.14	0.032	.8342871 .9918978
pyperstudlep		1.009294	.0019578	4.77	0.000	1.005464 1.013138
pyperstude~s		.9831259	.0022365	-7.48	0.000	.9787523 .987519
pyperstudw~e		.9910625	.0056214	-1.58	0.113	.9801058 1.002142
pyperstudb~k		.9934337	.0063514	-1.03	0.303	.9810629 1.00596
pyperstudl~o		.9980076	.0060001	-0.33	0.740	.9863167 1.009837
pyperstudm~e		1.013906	.0040996	3.42	0.001	1.005903 1.021973
male		1.599915	.1395654	5.39	0.000	1.348479 1.898234
age		.9777939	.0043412	-5.06	0.000	.9693222 .9863397
Asian		.8773283	.4164942	-0.28	0.783	.345996 2.224606
Black		1.357933	.1356265	3.06	0.002	1.11651 1.651558
Latino		1.073166	.0900031	0.84	0.400	.9104989 1.264895
Other		.9467153	.2805456	-0.18	0.853	.5296365 1.692236
AsianMale		2.23444	1.811479	0.99	0.321	.456134 10.94573
BlackMale		.8243359	.1692292	-0.94	0.347	.5512641 1.232675

LatinoMale		.8262679	.1267443	-1.24	0.213	.6117188	1.116066
OtherMale		1.339408	.8317443	0.47	0.638	.3965835	4.523673
-----							
3							
tenure		1.019346	.0036423	5.36	0.000	1.012232	1.02651
exper		1.070123	.0055465	13.08	0.000	1.059307	1.081049
acctrating2		1.029302	.0416212	0.71	0.475	.9508746	1.114198
pyperstudlep		.9987876	.0018443	-0.66	0.511	.9951793	1.002409
pyperstude~s		1.002754	.0020732	1.33	0.183	.9986985	1.006826
pyperstudw~e		1.010859	.0059816	1.83	0.068	.9992031	1.022651
pyperstudb~k		1.00142	.0065897	0.22	0.829	.9885871	1.014419
pyperstudl~o		1.002676	.0062505	0.43	0.668	.9904998	1.015002
pyperstudm~e		1.002198	.0039222	0.56	0.575	.9945399	1.009915
male		2.524627	.1925404	12.14	0.000	2.174105	2.931663
age		.9798189	.0043306	-4.61	0.000	.9713678	.9883436
Asian		1.10841	.5400499	0.21	0.833	.4265501	2.880254
Black		.9547196	.0961715	-0.46	0.646	.7836678	1.163107
Latino		1.204155	.0967454	2.31	0.021	1.028713	1.409517
Other		.5518079	.180828	-1.81	0.070	.2903008	1.048884
AsianMale		.4712751	.4432626	-0.80	0.424	.0745879	2.977698
BlackMale		.6111517	.1221218	-2.46	0.014	.4131044	.9041453
LatinoMale		.6006992	.0864704	-3.54	0.000	.45303	.7965024
OtherMale		.4594	.3682091	-0.97	0.332	.0954889	2.210187
-----							

-----> pyschooltype = M

Iteration 0: log likelihood = -5820.3664  
Iteration 1: log likelihood = -5525.2007  
Iteration 2: log likelihood = -5519.7526  
Iteration 3: log likelihood = -5519.6919  
Iteration 4: log likelihood = -5519.6917

Multinomial logistic regression	Number of obs	=	5543
	LR chi2(38)	=	601.35
	Prob > chi2	=	0.0000
Log likelihood = -5519.6917	Pseudo R2	=	0.0517

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----							
1		(base outcome)					
-----							
2							
tenure		1.011756	.0050787	2.33	0.020	1.001851	1.021759
exper		1.00626	.0062651	1.00	0.316	.9940558	1.018615
acctrating2		.8939641	.063276	-1.58	0.113	.7781636	1.026997
pyperstudlep		1.009872	.0042123	2.36	0.019	1.00165	1.018162
pyperstude~s		.992811	.0028007	-2.56	0.011	.9873369	.9983154
pyperstudw~e		.9983295	.0079429	-0.21	0.834	.9828823	1.014019
pyperstudb~k		1.002287	.0088665	0.26	0.796	.9850588	1.019816
pyperstudl~o		.9992713	.008464	-0.09	0.931	.9828191	1.015999
pyperstudm~e		1.01287	.0045854	2.82	0.005	1.003923	1.021898
male		1.477094	.1177311	4.89	0.000	1.263465	1.726842
age		.9649864	.0049729	-6.92	0.000	.9552888	.9747825
Asian		.6122836	.2994521	-1.00	0.316	.234775	1.59681
Black		1.001341	.1249533	0.01	0.991	.7840864	1.278792
Latino		1.72082	.2129804	4.39	0.000	1.350159	2.193239
Other		.4962216	.2329875	-1.49	0.136	.1977046	1.245474
AsianMale		1.514477	1.172721	0.54	0.592	.3320038	6.908478
BlackMale		1.690381	.3202441	2.77	0.006	1.166067	2.450448
LatinoMale		1.019051	.1693882	0.11	0.910	.7357119	1.411509
OtherMale		2.004569	1.217191	1.15	0.252	.6097676	6.589883
-----							
3							
tenure		1.014986	.0051146	2.95	0.003	1.005011	1.02506

exper		1.065605	.0078657	8.61	0.000	1.050299	1.081133
acctrating2		1.061707	.0831404	0.76	0.444	.9106443	1.23783
pyperstudlep		1.007448	.0051443	1.45	0.146	.9974159	1.017581
pyperstude~s		1.01164	.0032809	3.57	0.000	1.00523	1.018091
pyperstudw~e		1.022441	.0104111	2.18	0.029	1.002238	1.043052
pyperstudb~k		1.002616	.0113648	0.23	0.818	.9805871	1.02514
pyperstudl~o		1.002632	.0108423	0.24	0.808	.9816048	1.024109
pyperstudm~e		.9993107	.0059434	-0.12	0.908	.9877294	1.011028
male		1.797605	.1579726	6.67	0.000	1.513182	2.13549
age		.9641664	.0063865	-5.51	0.000	.9517301	.9767653
Asian		.5568112	.4239885	-0.77	0.442	.1251853	2.476637
Black		.9093768	.1474686	-0.59	0.558	.6617733	1.249622
Latino		1.495991	.2332701	2.58	0.010	1.102049	2.030753
Other		.1643124	.169257	-1.75	0.080	.02182	1.237333
AsianMale		1.591274	1.656553	0.45	0.655	.2068353	12.24236
BlackMale		1.988456	.450465	3.03	0.002	1.275512	3.099896
LatinoMale		.9979458	.198596	-0.01	0.992	.6756375	1.474009
OtherMale		6.133806	6.843554	1.63	0.104	.6887081	54.6292

-----> pyschooltype = S

```
Iteration 0: log likelihood = -6588.2959
Iteration 1: log likelihood = -6299.4774
Iteration 2: log likelihood = -6288.9998
Iteration 3: log likelihood = -6288.7516
Iteration 4: log likelihood = -6288.696
Iteration 5: log likelihood = -6288.6829
Iteration 6: log likelihood = -6288.6802
Iteration 7: log likelihood = -6288.6797
Iteration 8: log likelihood = -6288.6797
Iteration 9: log likelihood = -6288.6796
```

Multinomial logistic regression	Number of obs	=	6874
	LR chi2(38)	=	599.23
	Prob > chi2	=	0.0000
Log likelihood = -6288.6796	Pseudo R2	=	0.0455

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.005137	.0039511	1.30	0.192	.9974226 1.012911
exper		1.009471	.0054023	1.76	0.078	.9989382 1.020115
acctrating2		.9019339	.0607266	-1.53	0.125	.7904308 1.029166
pyperstudlep		1.021862	.0048179	4.59	0.000	1.012463 1.031349
pyperstude~s		.9857389	.0023914	-5.92	0.000	.981063 .990437
pyperstudw~e		1.001344	.0068591	0.20	0.845	.9879902 1.014878
pyperstudb~k		1.005403	.0075907	0.71	0.475	.9906355 1.020391
pyperstudl~o		1.006571	.0073424	0.90	0.369	.9922827 1.021065
pyperstudm~e		1.008792	.0036254	2.44	0.015	1.001712 1.015923
male		1.174807	.0810598	2.33	0.020	1.026207 1.344925
age		.9716824	.0046661	-5.98	0.000	.9625799 .980871
Asian		.5278113	.343947	-0.98	0.327	.1471618 1.89305
Black		1.292817	.1593124	2.08	0.037	1.015417 1.646
Latino		1.608428	.1957945	3.90	0.000	1.267023 2.041825
Other		.4498124	.2514519	-1.43	0.153	.1503823 1.345445
AsianMale		1.791263	1.538819	0.68	0.497	.3325968 9.647184
BlackMale		2.271665	.3849438	4.84	0.000	1.629688 3.166534
LatinoMale		1.226855	.1822291	1.38	0.169	.9169832 1.641441
OtherMale		3.281067	2.234184	1.74	0.081	.8637828 12.46309
3						
tenure		.9942578	.004703	-1.22	0.223	.9850828 1.003518

exper		1.051019	.0076595	6.83	0.000	1.036113	1.066139
acctrating2		1.19671	.1020427	2.11	0.035	1.012529	1.414394
pyperstudlep		.9986123	.0072672	-0.19	0.849	.9844699	1.012958
pyperstude~s		1.016116	.0031507	5.16	0.000	1.00996	1.022311
pyperstudw~e		1.059456	.0137778	4.44	0.000	1.032793	1.086807
pyperstudb~k		1.04702	.0145433	3.31	0.001	1.0189	1.075916
pyperstudl~o		1.043969	.0141126	3.18	0.001	1.016672	1.071999
pyperstudm~e		.995206	.0054172	-0.88	0.377	.984645	1.00588
male		1.877312	.1784969	6.62	0.000	1.558129	2.261881
age		.9784678	.006808	-3.13	0.002	.9652149	.9919026
Asian		.6505576	.6833723	-0.41	0.682	.0830135	5.098269
Black		.6540578	.1597647	-1.74	0.082	.4052246	1.05569
Latino		1.320739	.251778	1.46	0.144	.9089684	1.919045
Other		.8767661	.6642689	-0.17	0.862	.1986019	3.870651
AsianMale		1.42e-06	.0009555	-0.02	0.984	0	.
BlackMale		3.104652	.8897798	3.95	0.000	1.770363	5.444571
LatinoMale		.8187726	.1821659	-0.90	0.369	.5293981	1.266322
OtherMale		1.653576	1.522178	0.55	0.585	.2721832	10.04585

#### 2004-05

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperst
> udmobile Asian Black Latino Other AsianMale BlackMale LatinoMale OtherMale if
validcert2005a==1, rrr
```

```
Iteration 0: log likelihood = -24700.874
Iteration 1: log likelihood = -23590.164
Iteration 2: log likelihood = -23568.37
Iteration 3: log likelihood = -23568.323
Iteration 4: log likelihood = -23568.323
```

```
Multinomial logistic regression      Number of obs   =      23258
LR chi2(38)                        =      2265.10
Prob > chi2                         =      0.0000
Log likelihood = -23568.323          Pseudo R2       =      0.0459
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.010831	.0024966	4.36	0.000	1.005949 1.015736
exper		1.005706	.0031865	1.80	0.073	.9994802 1.011971
male		1.216823	.0504557	4.73	0.000	1.121843 1.319844
age		.9732073	.0025706	-10.28	0.000	.9681821 .9782586
pyacctrati~2		.950158	.0275134	-1.77	0.077	.8977345 1.005643
pyperstudlep		1.007439	.0013255	5.63	0.000	1.004845 1.010041
pyperstude~s		.9902531	.001225	-7.92	0.000	.9878551 .9926569
pyperstudw~e		.9929874	.0035999	-1.94	0.052	.9859568 1.000068
pyperstudb~k		.9977806	.0040202	-0.55	0.581	.9899322 1.005691
pyperstudl~o		.9980885	.0038144	-0.50	0.617	.9906403 1.005593
pyperstudm~e		1.004372	.0020674	2.12	0.034	1.000328 1.008433
Asian		.9734874	.2514702	-0.10	0.917	.5867443 1.615146
Black		1.149569	.0713464	2.25	0.025	1.017903 1.298266
Latino		1.34557	.0767681	5.20	0.000	1.203215 1.504768
Other		.7238426	.1509079	-1.55	0.121	.4810434 1.089191
AsianMale		1.45982	.6514127	0.85	0.397	.6087926 3.500494
BlackMale		1.523129	.1508264	4.25	0.000	1.254432 1.84938
LatinoMale		1.124048	.0927311	1.42	0.156	.9562315 1.321317
OtherMale		1.672711	.5195042	1.66	0.098	.9100345 3.074566
-----						
3						
tenure		1.012496	.0023509	5.35	0.000	1.007899 1.017114
exper		1.057949	.0036201	16.46	0.000	1.050877 1.065068

male		1.398009	.0589981	7.94	0.000	1.287028	1.51856
age		.9785514	.0029736	-7.14	0.000	.9727407	.9843969
pyacctrati~2		1.378261	.0399695	11.06	0.000	1.302107	1.458869
pyperstudlep		1.00993	.0014071	7.09	0.000	1.007175	1.012691
pyperstude~s		1.01934	.0013528	14.43	0.000	1.016692	1.021995
pyperstudw~e		1.010245	.0042334	2.43	0.015	1.001981	1.018576
pyperstudb~k		.9963817	.0046443	-0.78	0.437	.9873204	1.005526
pyperstudl~o		.991927	.0043816	-1.84	0.067	.9833763	1.000552
pyperstudm~e		.9901957	.0024822	-3.93	0.000	.9853426	.9950727
Asian		.4883856	.1841987	-1.90	0.057	.2331981	1.022823
Black		.730958	.0541092	-4.23	0.000	.6322404	.8450893
Latino		1.167492	.0731911	2.47	0.014	1.032503	1.320129
Other		.5132891	.1318326	-2.60	0.009	.3102705	.8491484
AsianMale		1.390346	.8703409	0.53	0.599	.4076434	4.742043
BlackMale		1.623271	.1870942	4.20	0.000	1.295042	2.034691
LatinoMale		.9424294	.0860987	-0.65	0.516	.7879243	1.127231
OtherMale		1.841173	.6869488	1.64	0.102	.8861516	3.825438

-----

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=23258)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		35.234	2	0.000
exper		300.098	2	0.000
male		67.675	2	0.000
age		126.571	2	0.000
pyacctrati~2		158.767	2	0.000
pyperstudlep		60.587	2	0.000
pyperstude~s		401.086	2	0.000
pyperstudw~e		15.028	2	0.001
pyperstudb~k		0.695	2	0.706
pyperstudl~o		3.345	2	0.188
pyperstudm~e		29.141	2	0.000
Asian		4.369	2	0.113
Black		32.532	2	0.000
Latino		27.459	2	0.000
Other		8.297	2	0.016
AsianMale		0.805	2	0.669
BlackMale		26.156	2	0.000
LatinoMale		3.716	2	0.156
OtherMale		4.253	2	0.119

-----

\*\*\*\* Wald tests for independent variables (N=23258)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		35.213	2	0.000
exper		282.305	2	0.000
male		67.538	2	0.000
age		124.615	2	0.000
pyacctrati~2		159.844	2	0.000
pyperstudlep		60.024	2	0.000
pyperstude~s		387.446	2	0.000
pyperstudw~e		14.819	2	0.001
pyperstudb~k		0.695	2	0.706
pyperstudl~o		3.389	2	0.184
pyperstudm~e		27.996	2	0.000
Asian		3.736	2	0.154
Black		31.795	2	0.000

Latino		27.499	2	0.000
Other		7.761	2	0.021
AsianMale		0.799	2	0.671
BlackMale		25.831	2	0.000
LatinoMale		3.717	2	0.156
OtherMale		4.211	2	0.122

\*\*\*\* Hausman tests of IIA assumption (N=23258)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-83.182	20	---	---
3		-5.550	19	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=23258)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-4954.014	-4939.784	28.460	20	0.099	for Ho
3		-5650.783	-5645.361	10.844	20	0.950	for Ho

\*\*\*\* Wald tests for combining alternatives (N=23258)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1494.467	19	0.000
2-	1	458.601	19	0.000
3-	1	1292.120	19	0.000

\*\*\*\* LR tests for combining alternatives (N=23258)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1669.276	19	0.000
2-	1	473.234	19	0.000
3-	1	1428.344	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstuddecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
> AsianMale BlackMale LatinoMale OtherMale if validcert2005a==1, rrr
```

-----> region = 1

```
Iteration 0: log likelihood = -1740.9362
Iteration 1: log likelihood = -1651.521
Iteration 2: log likelihood = -1648.0365
Iteration 3: log likelihood = -1647.7261
Iteration 4: log likelihood = -1647.6834
```

```

Iteration 5: log likelihood = -1647.6734
Iteration 6: log likelihood = -1647.671
Iteration 7: log likelihood = -1647.6705
Iteration 8: log likelihood = -1647.6704
Iteration 9: log likelihood = -1647.6704
Iteration 10: log likelihood = -1647.6704

```

```

Multinomial logistic regression
Log likelihood = -1647.6704
Number of obs   = 1618
LR chi2(38)    = 186.53
Prob > chi2    = 0.0000
Pseudo R2     = 0.0536

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.018743	.0090756	2.08	0.037	1.001109	1.036687
	exper	.9921435	.0128197	-0.61	0.542	.9673328	1.017591
	male	1.014762	.3045054	0.05	0.961	.5635587	1.827212
	age	.9809635	.0108414	-1.74	0.082	.9599432	1.002444
pyacctrati~2		.7980171	.1005423	-1.79	0.073	.6234034	1.02154
pyperstudlep		1.00611	.0031377	1.95	0.051	.9999788	1.012278
pyperstude~s		.9936644	.0059601	-1.06	0.289	.9820512	1.005415
pyperstudw~e		.9748414	.0749057	-0.33	0.740	.8385493	1.133285
pyperstudb~k		.8695447	.1313314	-0.93	0.355	.6467418	1.169103
pyperstudl~o		.9805592	.0656355	-0.29	0.769	.8599973	1.118022
pyperstudm~e		.9930867	.0083293	-0.83	0.408	.976895	1.009547
	Asian	8.09e-08	.0001889	-0.01	0.994	0	.
	Black	.9151426	6927.706	-0.00	1.000	0	.
	Latino	.9080522	.1922225	-0.46	0.649	.5996845	1.374987
	Other	.8125912	.7625241	-0.22	0.825	.1291566	5.112433
	AsianMale	1.31e+07	3.05e+10	0.01	0.994	0	.
	BlackMale	.4267678	3230.668	-0.00	1.000	0	.
	LatinoMale	1.302929	.4284102	0.80	0.421	.6839763	2.481993
	OtherMale	1.421947	10764.26	0.00	1.000	0	.
3							
	tenure	1.030004	.0095945	3.17	0.002	1.01137	1.048981
	exper	1.086692	.0186479	4.84	0.000	1.050751	1.123863
	male	.7061247	.2489045	-0.99	0.324	.3538655	1.409044
	age	.9451947	.0148113	-3.60	0.000	.9166063	.9746748
pyacctrati~2		1.419569	.1962437	2.53	0.011	1.082642	1.86135
pyperstudlep		1.017609	.0036018	4.93	0.000	1.010574	1.024693
pyperstude~s		.9967084	.0074553	-0.44	0.659	.9822028	1.011428
pyperstudw~e		1.298923	.14096	2.41	0.016	1.050052	1.60678
pyperstudb~k		1.280714	.137904	2.30	0.022	1.037044	1.581639
pyperstudl~o		1.26914	.1239915	2.44	0.015	1.047971	1.536985
pyperstudm~e		1.000019	.0089443	0.00	0.998	.9826416	1.017704
	Asian	1.73e-07	.0004562	-0.01	0.995	0	.
	Black	8.11e+07	4.18e+11	0.00	0.997	0	.
	Latino	.6511751	.1512268	-1.85	0.065	.4130641	1.026545
	Other	.453868	.5482424	-0.65	0.513	.0425338	4.843112
	AsianMale	8.802396	29599.5	0.00	0.999	0	.
	BlackMale	3.28e-15	1.74e-11	-0.01	0.995	0	.
	LatinoMale	1.930963	.7461186	1.70	0.089	.9054659	4.1179
	OtherMale	3.64e+08	1.88e+12	0.00	0.997	0	.

```

-----> region = 2

```

```

Iteration 0: log likelihood = -739.56901
Iteration 1: log likelihood = -689.40693
Iteration 2: log likelihood = -686.98457
Iteration 3: log likelihood = -686.68311

```



```

Iteration 4: log likelihood = -686.63527
Iteration 5: log likelihood = -686.62404
Iteration 6: log likelihood = -686.62152
Iteration 7: log likelihood = -686.62097
Iteration 8: log likelihood = -686.62088
Iteration 9: log likelihood = -686.62087

```

```

Multinomial logistic regression
Number of obs   =      733
LR chi2(38)     =     105.90
Prob > chi2     =      0.0000
Pseudo R2      =      0.0716

Log likelihood = -686.62087

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.019079	.0170375	1.13	0.258	.9862278	1.053025
	exper	.9871295	.0202842	-0.63	0.528	.9481631	1.027697
	male	1.959811	.5926096	2.23	0.026	1.083495	3.544879
	age	.9581159	.0160076	-2.56	0.010	.9272497	.9900096
pyacctrati~2		.818762	.1707869	-0.96	0.338	.5440067	1.232285
pyperstudlep		.9721552	.0160673	-1.71	0.088	.9411685	1.004162
pyperstude~s		1.008668	.0075858	1.15	0.251	.9939088	1.023646
pyperstudw~e		.9830388	.0835931	-0.20	0.841	.8321247	1.161323
pyperstudb~k		.9565206	.0919032	-0.46	0.644	.7923376	1.154725
pyperstudl~o		.9766327	.079835	-0.29	0.772	.8320503	1.146339
pyperstudm~e		1.014254	.0134169	1.07	0.285	.988295	1.040894
	Asian	6.14e-07	.0021065	-0.00	0.997	0	.
	Black	3.067711	2.195923	1.57	0.117	.7542392	12.47728
	Latino	1.624398	.4533992	1.74	0.082	.9399538	2.807232
	Other	.7781309	.9005738	-0.22	0.828	.0805212	7.519606
	AsianMale	.3493147	1465.659	-0.00	1.000	0	.
	BlackMale	.7510074	.9487518	-0.23	0.821	.0631418	8.932475
	LatinoMale	.6001239	.2479128	-1.24	0.216	.2670591	1.348573
	OtherMale	1.387762	2.330372	0.20	0.845	.0516358	37.29742
3							
	tenure	.9885095	.0125243	-0.91	0.362	.9642645	1.013364
	exper	1.052887	.0183182	2.96	0.003	1.017589	1.089409
	male	.8698132	.2509291	-0.48	0.629	.4941583	1.531038
	age	.9923394	.0157334	-0.49	0.628	.9619766	1.02366
pyacctrati~2		1.531138	.2685839	2.43	0.015	1.085679	2.159373
pyperstudlep		.9984496	.013968	-0.11	0.912	.9714449	1.026205
pyperstude~s		1.023282	.0077358	3.04	0.002	1.008232	1.038556
pyperstudw~e		1.147627	.0976863	1.62	0.106	.9712843	1.355986
pyperstudb~k		1.163231	.1101139	1.60	0.110	.9662497	1.400369
pyperstudl~o		1.126236	.0924915	1.45	0.148	.9587931	1.322921
pyperstudm~e		.9618233	.0156052	-2.40	0.016	.9317189	.9929004
	Asian	8.10e-07	.0025336	-0.00	0.996	0	.
	Black	1.598759	1.152062	0.65	0.515	.3894118	6.563825
	Latino	.8192096	.2060629	-0.79	0.428	.5003627	1.341236
	Other	4.63e-07	.0005619	-0.01	0.990	0	.
	AsianMale	.4719602	1798.912	-0.00	1.000	0	.
	BlackMale	.7415433	1.083083	-0.20	0.838	.0423535	12.98326
	LatinoMale	1.704239	.6794575	1.34	0.181	.7801312	3.723002
	OtherMale	1.269266	2313.323	0.00	1.000	0	.

```
-----> region = 3
```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -327.52666

```

```

Iteration 1: log likelihood = -295.45406
Iteration 2: log likelihood = -294.30385
Iteration 3: log likelihood = -294.16588
Iteration 4: log likelihood = -294.14393
Iteration 5: log likelihood = -294.13876
Iteration 6: log likelihood = -294.13755
Iteration 7: log likelihood = -294.13729
Iteration 8: log likelihood = -294.13723
Iteration 9: log likelihood = -294.13722

```

```

Multinomial logistic regression      Number of obs   =      304
                                     LR chi2(32)        =      66.78
                                     Prob > chi2         =      0.0003
Log likelihood = -294.13722          Pseudo R2        =      0.1019

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.018446	.0253576	0.73	0.463	.9699387	1.069378
	exper	.9413629	.0280851	-2.03	0.043	.8878957	.9980499
	male	2.236938	.8391839	2.15	0.032	1.072326	4.666389
	age	.9972038	.0255369	-0.11	0.913	.9483878	1.048532
pyacctrati~2		1.199612	.3347289	0.65	0.514	.694271	2.072776
pyperstudlep		1.051317	.0427725	1.23	0.219	.9707395	1.138582
pyperstude~s		.9825551	.0177925	-0.97	0.331	.948294	1.018054
pyperstudw~e		1.022325	.0613078	0.37	0.713	.9089572	1.149833
pyperstudb~k		1.043364	.0638404	0.69	0.488	.9254503	1.1763
pyperstudl~o		1.04802	.065693	0.75	0.454	.9268596	1.18502
pyperstudm~e		1.015195	.0237518	0.64	0.519	.9696939	1.062832
	Asian	(omitted)					
	Black	1.305006	1.003731	0.35	0.729	.2890147	5.892574
	Latino	2.835797	1.591789	1.86	0.063	.9437978	8.520622
	Other	1.354299	3003.564	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	1256814	7.85e+08	0.02	0.982	0	.
	LatinoMale	2.994443	3.703972	0.89	0.375	.2651084	33.82273
	OtherMale	(omitted)					
3							
	tenure	1.011385	.0205954	0.56	0.578	.9718133	1.052567
	exper	1.01676	.0292505	0.58	0.563	.9610167	1.075737
	male	1.504423	.5020508	1.22	0.221	.7821858	2.893543
	age	.9868924	.026035	-0.50	0.617	.9371615	1.039262
pyacctrati~2		.9304735	.223348	-0.30	0.764	.5812784	1.489443
pyperstudlep		1.050034	.0374038	1.37	0.171	.9792244	1.125964
pyperstude~s		1.03919	.0175723	2.27	0.023	1.005313	1.074208
pyperstudw~e		1.125979	.0681328	1.96	0.050	1.000056	1.267758
pyperstudb~k		1.091869	.0672442	1.43	0.154	.9677169	1.23195
pyperstudl~o		1.103158	.0694103	1.56	0.119	.97517	1.247944
pyperstudm~e		.9993824	.0242307	-0.03	0.980	.9530019	1.04802
	Asian	(omitted)					
	Black	1.437933	.9040657	0.58	0.563	.4193402	4.930728
	Latino	.8348121	.5318273	-0.28	0.777	.2395081	2.90976
	Other	5580878	7.59e+09	0.01	0.991	0	.
	AsianMale	(omitted)					
	BlackMale	427260.2	2.67e+08	0.02	0.983	0	.
	LatinoMale	10.4392	13.34962	1.83	0.067	.8514384	127.9916
	OtherMale	(omitted)					

```

-----> region = 4

```

```

Iteration 0: log likelihood = -5171.5823

```

```

Iteration 1: log likelihood = -4870.9237
Iteration 2: log likelihood = -4856.9663
Iteration 3: log likelihood = -4856.8189
Iteration 4: log likelihood = -4856.7876
Iteration 5: log likelihood = -4856.7801
Iteration 6: log likelihood = -4856.7786
Iteration 7: log likelihood = -4856.7782
Iteration 8: log likelihood = -4856.7782
Iteration 9: log likelihood = -4856.7781

```

```

Multinomial logistic regression      Number of obs   =      5071
                                     LR chi2(38)       =      629.61
                                     Prob > chi2       =      0.0000
Log likelihood = -4856.7781         Pseudo R2       =      0.0609

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.019611	.0054352	3.64	0.000	1.009014	1.03032
	exper	1.034187	.0074577	4.66	0.000	1.019673	1.048908
	male	1.093746	.0990581	0.99	0.322	.9158518	1.306195
	age	.9553555	.005662	-7.71	0.000	.9443224	.9665176
pyacctrati~2		.9897144	.0583527	-0.18	0.861	.881706	1.110954
pyperstudlep		.9969527	.003059	-0.99	0.320	.9909751	1.002966
pyperstude~s		.9927365	.0031718	-2.28	0.023	.9865393	.9989726
pyperstudw~e		.9980092	.0055701	-0.36	0.721	.9871515	1.008986
pyperstudb~k		1.000495	.006776	0.07	0.942	.9873025	1.013865
pyperstudl~o		1.005964	.0065529	0.91	0.361	.9932026	1.01889
pyperstudm~e		1.004615	.0050486	0.92	0.360	.9947686	1.014559
	Asian	1.381154	.5050165	0.88	0.377	.6745309	2.82802
	Black	.8297887	.0874288	-1.77	0.077	.6749671	1.020123
	Latino	1.395253	.1846826	2.52	0.012	1.076424	1.808515
	Other	.7134327	.2719081	-0.89	0.376	.3380163	1.505804
	AsianMale	1.281598	1.156255	0.28	0.783	.2186749	7.511114
	BlackMale	1.864548	.3095931	3.75	0.000	1.3466	2.581718
	LatinoMale	1.11509	.2409697	0.50	0.614	.730073	1.703153
	OtherMale	.4866417	.4245082	-0.83	0.409	.0880415	2.689868
3							
	tenure	1.036425	.006075	6.10	0.000	1.024587	1.048401
	exper	1.073041	.0095303	7.94	0.000	1.054524	1.091884
	male	1.084459	.1183287	0.74	0.457	.8756605	1.343044
	age	.9726128	.0073868	-3.66	0.000	.9582421	.987199
pyacctrati~2		1.275487	.0879521	3.53	0.000	1.114245	1.460062
pyperstudlep		1.018286	.0037732	4.89	0.000	1.010918	1.025709
pyperstude~s		1.018091	.0042658	4.28	0.000	1.009764	1.026486
pyperstudw~e		1.017843	.0072984	2.47	0.014	1.003638	1.032248
pyperstudb~k		1.00666	.0089164	0.75	0.454	.9893353	1.024289
pyperstudl~o		.9925193	.0083568	-0.89	0.372	.9762746	1.009034
pyperstudm~e		.9889396	.0073257	-1.50	0.133	.9746853	1.003402
	Asian	.4375726	.2900477	-1.25	0.212	.119351	1.604259
	Black	.4434752	.0625806	-5.76	0.000	.3363201	.5847711
	Latino	1.195376	.1910571	1.12	0.264	.87389	1.63513
	Other	.4962456	.2750353	-1.26	0.206	.1674686	1.470483
	AsianMale	5.64e-06	.0031367	-0.02	0.983	0	.
	BlackMale	2.283385	.4923461	3.83	0.000	1.49638	3.484305
	LatinoMale	1.388663	.3748778	1.22	0.224	.8181069	2.357131
	OtherMale	3.257914	2.874614	1.34	0.181	.5779415	18.36519

```

-----> region = 5

```

```

note: OtherMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -525.36986
Iteration 1: log likelihood = -484.1022
Iteration 2: log likelihood = -483.03544
Iteration 3: log likelihood = -482.85527
Iteration 4: log likelihood = -482.82106
Iteration 5: log likelihood = -482.81298
Iteration 6: log likelihood = -482.81157
Iteration 7: log likelihood = -482.81124
Iteration 8: log likelihood = -482.81116
Iteration 9: log likelihood = -482.81115

```

```

Multinomial logistic regression      Number of obs   =      506
                                     LR chi2(36)        =      85.12
                                     Prob > chi2         =      0.0000
Log likelihood = -482.81115          Pseudo R2        =      0.0810

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.000341	.0165994	0.02	0.984	.96833	1.03341
	exper	1.025645	.0259305	1.00	0.317	.9760609	1.077748
	male	3.145236	.9113412	3.95	0.000	1.782441	5.549979
	age	.9540187	.0210867	-2.13	0.033	.9135719	.9962561
pyacctrati~2		1.187139	.2541964	0.80	0.423	.7802575	1.806198
pyperstudlep		.9692014	.0358298	-0.85	0.397	.90146	1.042033
pyperstude~s		1.007659	.0101369	0.76	0.448	.9879853	1.027724
pyperstudw~e		.9891163	.0475887	-0.23	0.820	.9001068	1.086928
pyperstudb~k		.9881279	.0494967	-0.24	0.812	.8957261	1.090062
pyperstudl~o		1.000148	.0612829	0.00	0.998	.8869677	1.12777
pyperstudm~e		.9832091	.029008	-0.57	0.566	.927967	1.04174
	Asian	6.14e-07	.0010112	-0.01	0.993	0	.
	Black	1.999076	.8105741	1.71	0.088	.903002	4.425575
	Latino	1.792641	2.263234	0.46	0.644	.1509522	21.28861
	Other	1.03e-06	.0016918	-0.01	0.993	0	.
	AsianMale	1525358	2.51e+09	0.01	0.993	0	.
	BlackMale	.9517595	.5334473	-0.09	0.930	.3172789	2.855047
	LatinoMale	2.414934	4.144087	0.51	0.607	.0836031	69.75708
	OtherMale	(omitted)					
3							
	tenure	.9996267	.0147393	-0.03	0.980	.9711516	1.028937
	exper	1.058466	.0259342	2.32	0.020	1.008837	1.110537
	male	2.223987	.6164423	2.88	0.004	1.291805	3.828841
	age	.9799299	.0213183	-0.93	0.351	.939025	1.022617
pyacctrati~2		1.26405	.2503962	1.18	0.237	.8573323	1.863713
pyperstudlep		.9958532	.0352655	-0.12	0.907	.9290781	1.067428
pyperstude~s		1.033896	.0102023	3.38	0.001	1.014092	1.054087
pyperstudw~e		.9781701	.0428744	-0.50	0.615	.8976461	1.065917
pyperstudb~k		.9526947	.0436166	-1.06	0.290	.870931	1.042135
pyperstudl~o		.9524329	.0550362	-0.84	0.399	.850448	1.066648
pyperstudm~e		.9674013	.0289016	-1.11	0.267	.9123818	1.025739
	Asian	5.77e-07	.0009341	-0.01	0.993	0	.
	Black	2.205418	.9388428	1.86	0.063	.9574934	5.079793
	Latino	2.508912	3.315736	0.70	0.486	.1881706	33.45176
	Other	2.77e-06	.0044826	-0.01	0.994	0	.
	AsianMale	3.073815	5867.783	0.00	1.000	0	.
	BlackMale	1.568208	.9240536	0.76	0.445	.4941287	4.976994
	LatinoMale	8.78e-07	.0008494	-0.01	0.989	0	.
	OtherMale	(omitted)					

```

-----> region = 6

```

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -851.34288  
 Iteration 1: log likelihood = -801.53306  
 Iteration 2: log likelihood = -799.99281  
 Iteration 3: log likelihood = -799.86746  
 Iteration 4: log likelihood = -799.84816  
 Iteration 5: log likelihood = -799.84353  
 Iteration 6: log likelihood = -799.84255  
 Iteration 7: log likelihood = -799.84234  
 Iteration 8: log likelihood = -799.8423  
 Iteration 9: log likelihood = -799.84229

Multinomial logistic regression	Number of obs	=	797
	LR chi2(36)	=	103.00
	Prob > chi2	=	0.0000
Log likelihood = -799.84229	Pseudo R2	=	0.0605

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	.9994508	.0133506	-0.04	0.967	.9736237 1.025963
exper	1.020409	.0168367	1.22	0.221	.9879378 1.053948
male	1.42729	.2793252	1.82	0.069	.9725889 2.094571
age	.9703984	.0136767	-2.13	0.033	.9439595 .9975779
pyacctrati~2	1.051362	.149899	0.35	0.725	.7950444 1.390315
pyperstudlep	1.040345	.0175742	2.34	0.019	1.006464 1.075366
pyperstude~s	.98448	.0071709	-2.15	0.032	.9705251 .9986355
pyperstudw~e	1.035225	.0329073	1.09	0.276	.972696 1.101774
pyperstudb~k	1.049126	.0360617	1.40	0.163	.9807749 1.122241
pyperstudl~o	1.030302	.0335471	0.92	0.359	.9666047 1.098196
pyperstudm~e	1.002858	.0131832	0.22	0.828	.9773496 1.029033
Asian	5.14e-07	.0009937	-0.01	0.994	0 .
Black	.7191077	.3674511	-0.65	0.519	.2641467 1.957684
Latino	2.652117	1.622929	1.59	0.111	.7993019 8.799834
Other	1.06e-06	.0020481	-0.01	0.994	0 .
AsianMale	(omitted)				
BlackMale	3.268831	2.512418	1.54	0.123	.7247034 14.74432
LatinoMale	.3972035	.3467658	-1.06	0.290	.0717627 2.198505
OtherMale	1.00e+12	2.19e+15	0.01	0.990	0 .
3					
tenure	.9830818	.0128325	-1.31	0.191	.9582495 1.008558
exper	1.100904	.0217057	4.88	0.000	1.059173 1.144279
male	2.215185	.4588301	3.84	0.000	1.47605 3.324443
age	.9396294	.0165058	-3.54	0.000	.9078291 .9725436
pyacctrati~2	1.362885	.2101248	2.01	0.045	1.007451 1.843718
pyperstudlep	1.04848	.0187404	2.65	0.008	1.012385 1.085861
pyperstude~s	1.014535	.0077382	1.89	0.059	.999481 1.029815
pyperstudw~e	1.039645	.0335944	1.20	0.229	.9758432 1.107619
pyperstudb~k	1.055275	.0358183	1.59	0.113	.9873565 1.127865
pyperstudl~o	1.012159	.0333554	0.37	0.714	.9488499 1.079692
pyperstudm~e	.9932932	.0158502	-0.42	0.673	.9627082 1.02485
Asian	5.63e-07	.0011566	-0.01	0.994	0 .
Black	1.328636	.6117796	0.62	0.537	.5388459 3.276028
Latino	1.780469	1.303158	0.79	0.431	.424155 7.473848
Other	8.41e-07	.0017272	-0.01	0.995	0 .
AsianMale	(omitted)				
BlackMale	.3911314	.3529425	-1.04	0.298	.066716 2.293061
LatinoMale	.3286956	.3662337	-1.00	0.318	.0370154 2.918807
OtherMale	7.58e+11	1.73e+15	0.01	0.990	0 .

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1021.2652  
 Iteration 1: log likelihood = -955.80807  
 Iteration 2: log likelihood = -953.28472  
 Iteration 3: log likelihood = -953.12739  
 Iteration 4: log likelihood = -953.09222  
 Iteration 5: log likelihood = -953.08505  
 Iteration 6: log likelihood = -953.08374  
 Iteration 7: log likelihood = -953.08361  
 Iteration 8: log likelihood = -953.08358

Multinomial logistic regression	Number of obs	=	1000
	LR chi2(36)	=	136.36
	Prob > chi2	=	0.0000
Log likelihood = -953.08358	Pseudo R2	=	0.0668

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9892562	.0131529	-0.81	0.417	.9638099	1.015374
	exper	.9985379	.016265	-0.09	0.928	.9671625	1.030931
	male	1.146507	.2298061	0.68	0.495	.7740366	1.698211
	age	.9776539	.0140848	-1.57	0.117	.9504342	1.005653
pyacctrati~2		.7975222	.1361673	-1.33	0.185	.5707033	1.114487
pyperstudlep		.9794685	.0169242	-1.20	0.230	.9468531	1.013207
pyperstude~s		.988587	.0083133	-1.36	0.172	.9724267	1.005016
pyperstudw~e		.9136147	.0706515	-1.17	0.243	.7851238	1.063134
pyperstudb~k		.9171003	.0724791	-1.09	0.274	.7854992	1.07075
pyperstudl~o		.9448821	.0749256	-0.71	0.475	.8088734	1.10376
pyperstudm~e		.9965623	.014071	-0.24	0.807	.9693618	1.024526
	Asian	1.03e-06	.0015119	-0.01	0.993	0	.
	Black	1.85349	.6529128	1.75	0.080	.929276	3.696882
	Latino	1.461084	1.838222	0.30	0.763	.1240965	17.20247
	Other	3.06e-06	.0021181	-0.02	0.985	0	.
	AsianMale	(omitted)					
	BlackMale	2.276683	1.185015	1.58	0.114	.8208266	6.314712
	LatinoMale	.9538294	1.499489	-0.03	0.976	.0437845	20.77883
	OtherMale	.4873077	611.149	-0.00	1.000	0	.
3							
	tenure	.9926311	.0101433	-0.72	0.469	.9729484	1.012712
	exper	1.024113	.0138141	1.77	0.077	.9973928	1.051549
	male	2.0562	.3496534	4.24	0.000	1.473398	2.86953
	age	1.000429	.0122864	0.03	0.972	.9766352	1.024802
pyacctrati~2		1.476805	.1920861	3.00	0.003	1.144481	1.905626
pyperstudlep		1.019551	.0158706	1.24	0.214	.9889147	1.051136
pyperstude~s		1.031946	.0071884	4.51	0.000	1.017953	1.046132
pyperstudw~e		.9325297	.0732466	-0.89	0.374	.7994734	1.087731
pyperstudb~k		.9235646	.073719	-1.00	0.319	.789813	1.079966
pyperstudl~o		.9067262	.0730054	-1.22	0.224	.7743572	1.061722
pyperstudm~e		.993877	.0121297	-0.50	0.615	.9703853	1.017937
	Asian	2.49e-06	.0028485	-0.01	0.991	0	.
	Black	1.179331	.3994256	0.49	0.626	.6072164	2.290486
	Latino	1.419272	1.779489	0.28	0.780	.1215698	16.56935
	Other	1.499719	1.877008	0.32	0.746	.1290236	17.43214
	AsianMale	(omitted)					
	BlackMale	1.328555	.6963411	0.54	0.588	.4755929	3.711277
	LatinoMale	.9991212	1.51389	-0.00	1.000	.0512695	19.4705
	OtherMale	5.45e-07	.0004203	-0.02	0.985	0	.

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -388.60186  
 Iteration 1: log likelihood = -342.26164  
 Iteration 2: log likelihood = -339.77205  
 Iteration 3: log likelihood = -339.48278  
 Iteration 4: log likelihood = -339.42315  
 Iteration 5: log likelihood = -339.41004  
 Iteration 6: log likelihood = -339.40793  
 Iteration 7: log likelihood = -339.40741  
 Iteration 8: log likelihood = -339.4073  
 Iteration 9: log likelihood = -339.40728

Multinomial logistic regression

Number of obs = 383  
 LR chi2(36) = 98.39  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1266

Log likelihood = -339.40728

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.045983	.0272559	1.73	0.084	.9939033	1.100791
	exper	.9468449	.0274791	-1.88	0.060	.8944901	1.002264
	male	3.606557	1.290329	3.59	0.000	1.788775	7.271598
	age	1.000861	.0248388	0.03	0.972	.9533427	1.050747
pyacctrati~2		.8779597	.2660082	-0.43	0.668	.4848124	1.589921
pyperstudlep		.9908096	.043837	-0.21	0.835	.9085106	1.080564
pyperstude~s		.991355	.0142228	-0.61	0.545	.963867	1.019627
pyperstudw~e		.7705202	.1258832	-1.60	0.111	.5593962	1.061325
pyperstudb~k		.7913397	.130077	-1.42	0.155	.5733847	1.092144
pyperstudl~o		.779993	.1292834	-1.50	0.134	.5636435	1.079386
pyperstudm~e		1.036675	.0294172	1.27	0.204	.9805924	1.095965
	Asian	.7950754	2576.054	-0.00	1.000	0	.
	Black	3.372839	2.005728	2.04	0.041	1.051504	10.81882
	Latino	4978002	3.14e+09	0.02	0.981	0	.
	Other	3.23e-06	.0030447	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	.1180971	.1575281	-1.60	0.109	.0086464	1.61304
	LatinoMale	3.29e-13	4.43e-10	-0.02	0.983	0	.
	OtherMale	3.01e+11	4.18e+14	0.02	0.985	0	.
3							
	tenure	1.024942	.0188399	1.34	0.180	.988674	1.062541
	exper	1.009513	.0240357	0.40	0.691	.9634862	1.057739
	male	2.735236	.7653934	3.60	0.000	1.58054	4.733519
	age	1.008886	.0227343	0.39	0.695	.9652976	1.054443
pyacctrati~2		1.786105	.3858275	2.69	0.007	1.16959	2.727597
pyperstudlep		.9959008	.0335302	-0.12	0.903	.9323043	1.063835
pyperstude~s		1.036097	.0116577	3.15	0.002	1.013498	1.059199
pyperstudw~e		.8648207	.1170651	-1.07	0.283	.6632916	1.127581
pyperstudb~k		.8466045	.114574	-1.23	0.219	.6493584	1.103765
pyperstudl~o		.845365	.1153015	-1.23	0.218	.6470632	1.104439
pyperstudm~e		1.0017	.0287964	0.06	0.953	.9468202	1.05976
	Asian	971660.2	1.59e+09	0.01	0.993	0	.
	Black	2.289917	1.292402	1.47	0.142	.7575505	6.921942
	Latino	2695289	1.70e+09	0.02	0.981	0	.
	Other	1.620454	2.045654	0.38	0.702	.1364845	19.23934
	AsianMale	(omitted)					
	BlackMale	1.098633	1.032684	0.10	0.920	.1740795	6.933582
	LatinoMale	8.66e-07	.0005464	-0.02	0.982	0	.
	OtherMale	694055.1	7.06e+08	0.01	0.989	0	.

-----> region = 9

note: AsianMale omitted because of collinearity  
 note: LatinoMale omitted because of collinearity  
 Iteration 0: log likelihood = -218.4905  
 Iteration 1: log likelihood = -186.21918  
 Iteration 2: log likelihood = -183.62632  
 Iteration 3: log likelihood = -183.36082  
 Iteration 4: log likelihood = -183.30656  
 Iteration 5: log likelihood = -183.29401  
 Iteration 6: log likelihood = -183.2919  
 Iteration 7: log likelihood = -183.29167  
 Iteration 8: log likelihood = -183.29162  
 Iteration 9: log likelihood = -183.29161

Multinomial logistic regression	Number of obs	=	208
	LR chi2(34)	=	70.40
	Prob > chi2	=	0.0002
Log likelihood = -183.29161	Pseudo R2	=	0.1611

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1						
tenure	1.011161	.0267753	0.42	0.675	.960021	1.065025
exper	.9700141	.0344308	-0.86	0.391	.9048249	1.0399
male	.4539432	.1761025	-2.04	0.042	.2122202	.9709934
age	1.00943	.0300287	0.32	0.752	.9522582	1.070035
pyacctrati~2	.8026273	.2488986	-0.71	0.478	.4370712	1.473926
pyperstudlep	.9299732	.0661135	-1.02	0.307	.8090156	1.069015
pyperstude~s	.9849503	.0160559	-0.93	0.352	.9539787	1.016927
pyperstudw~e	.8454726	.0977179	-1.45	0.146	.6740925	1.060424
pyperstudb~k	.841375	.1055848	-1.38	0.169	.6579179	1.075988
pyperstudl~o	.8424559	.1026622	-1.41	0.159	.6634669	1.069732
pyperstudm~e	1.003539	.0294186	0.12	0.904	.9475051	1.062887
Asian	1.431653	25742.27	0.00	1.000	0	.
Black	1.83e-07	.0002043	-0.01	0.989	0	.
Latino	3.78e-07	.0008258	-0.01	0.995	0	.
Other	1.07e+07	4.09e+10	0.00	0.997	0	.
AsianMale	(omitted)					
BlackMale	2.75e+13	4.27e+16	0.02	0.984	0	.
LatinoMale	(omitted)					
OtherMale	2.35e-14	9.56e-11	-0.01	0.994	0	.
2						
tenure	.9784016	.0349415	-0.61	0.541	.9122593	1.049339
exper	.970925	.044411	-0.65	0.519	.8876687	1.06199
male	.3425812	.1636831	-2.24	0.025	.1342966	.8739007
age	.9723392	.0350001	-0.78	0.436	.9061042	1.043416
pyacctrati~2	.6006635	.2308414	-1.33	0.185	.282817	1.275725
pyperstudlep	.974548	.0444424	-0.57	0.572	.891222	1.065665
pyperstude~s	.9447616	.0220715	-2.43	0.015	.9024777	.9890265
pyperstudw~e	.6488791	.0890553	-3.15	0.002	.4958391	.8491547
pyperstudb~k	.6703349	.0968513	-2.77	0.006	.5050199	.8897646
pyperstudl~o	.694687	.1009869	-2.51	0.012	.522456	.923695
pyperstudm~e	1.007815	.0332225	0.24	0.813	.9447586	1.075079
Asian	8.39e+08	1.00e+13	0.00	0.999	0	.
Black	.2268488	.2832476	-1.19	0.235	.0196296	2.621576
Latino	4.58e-07	.0013659	-0.00	0.996	0	.
Other	.5085838	3476.326	-0.00	1.000	0	.
AsianMale	(omitted)					
BlackMale	1.97e+07	2.12e+10	0.02	0.988	0	.
LatinoMale	(omitted)					
OtherMale	1.274893	8714.287	0.00	1.000	0	.
3	(base outcome)					



-----> region = 10

Iteration 0: log likelihood = -3632.1295  
 Iteration 1: log likelihood = -3423.5344  
 Iteration 2: log likelihood = -3419.3076  
 Iteration 3: log likelihood = -3419.2919  
 Iteration 4: log likelihood = -3419.2919

Multinomial logistic regression	Number of obs	=	3392
	LR chi2(38)	=	425.68
	Prob > chi2	=	0.0000
Log likelihood = -3419.2919	Pseudo R2	=	0.0586

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.016393	.0065811	2.51	0.012	1.003575	1.029374
	exper	1.012516	.0085008	1.48	0.138	.9959907	1.029315
	male	1.122336	.1199543	1.08	0.280	.9102219	1.38388
	age	.9688996	.0066745	-4.59	0.000	.9559056	.9820701
pyacctrati~2		.9526564	.0702495	-0.66	0.511	.8244573	1.10079
pyperstudlep		1.005602	.0048686	1.15	0.249	.9961049	1.01519
pyperstude~s		.9833023	.0034198	-4.84	0.000	.9766224	.9900278
pyperstudw~e		.9938219	.0083073	-0.74	0.458	.9776725	1.010238
pyperstudb~k		.9984573	.0088827	-0.17	0.862	.9811985	1.01602
pyperstudl~o		1.002868	.008661	0.33	0.740	.9860358	1.019988
pyperstudm~e		1.009126	.0056967	1.61	0.108	.9980218	1.020353
	Asian	1.493947	.9208099	0.65	0.515	.446363	5.00014
	Black	1.446295	.1989048	2.68	0.007	1.104571	1.893739
	Latino	2.429792	.4592797	4.70	0.000	1.677548	3.519358
	Other	.5568759	.2663689	-1.22	0.221	.2180748	1.422038
	AsianMale	.5307003	.5039565	-0.67	0.505	.0825169	3.413155
	BlackMale	1.503692	.3265745	1.88	0.060	.9824118	2.301571
	LatinoMale	1.135137	.3458897	0.42	0.677	.6247082	2.062621
	OtherMale	3.378205	2.175333	1.89	0.059	.9562479	11.93443
3							
	tenure	1.020374	.0063983	3.22	0.001	1.007911	1.032992
	exper	1.073651	.0103595	7.37	0.000	1.053538	1.094148
	male	.9482301	.1089825	-0.46	0.644	.7569775	1.187803
	age	.9667608	.0081388	-4.02	0.000	.9509399	.982845
pyacctrati~2		1.311122	.0992363	3.58	0.000	1.130361	1.520788
pyperstudlep		1.020112	.0057161	3.55	0.000	1.00897	1.031377
pyperstude~s		1.008074	.0037401	2.17	0.030	1.00077	1.015431
pyperstudw~e		1.001253	.0087541	0.14	0.886	.984241	1.018558
pyperstudb~k		.9931319	.0095127	-0.72	0.472	.9746612	1.011953
pyperstudl~o		.9773406	.0089028	-2.52	0.012	.9600462	.9949466
pyperstudm~e		.9998277	.0067711	-0.03	0.980	.9866443	1.013187
	Asian	1.34208	.9964245	0.40	0.692	.3131885	5.751103
	Black	.7758017	.129104	-1.53	0.127	.5598854	1.074985
	Latino	2.255179	.4705676	3.90	0.000	1.498196	3.394638
	Other	.251714	.1637841	-2.12	0.034	.0703156	.9010799
	AsianMale	1.024518	1.096794	0.02	0.982	.1256809	8.351607
	BlackMale	2.68732	.6741098	3.94	0.000	1.643603	4.393816
	LatinoMale	1.80006	.6008779	1.76	0.078	.9357235	3.462793
	OtherMale	2.457194	2.3109	0.96	0.339	.3889698	15.52255

-----> region = 11

Iteration 0: log likelihood = -2573.0985

```

Iteration 1:  log likelihood = -2433.2752
Iteration 2:  log likelihood = -2429.8054
Iteration 3:  log likelihood = -2429.7941
Iteration 4:  log likelihood = -2429.7941

```

```

Multinomial logistic regression          Number of obs   =      2384
                                         LR chi2(38)      =      286.61
                                         Prob > chi2      =      0.0000
Log likelihood = -2429.7941             Pseudo R2       =      0.0557

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.000915	.0079558	0.12	0.908	.9854425	1.01663
	exper	.9992453	.0097204	-0.08	0.938	.9803741	1.01848
	male	1.319464	.1517441	2.41	0.016	1.053187	1.653063
	age	.989248	.0078886	-1.36	0.175	.9739068	1.004831
	pyacctrati~2	1.040763	.0948858	0.44	0.661	.8704585	1.244387
	pyperstudlep	.9963819	.0069764	-0.52	0.605	.9828017	1.01015
	pyperstude~s	.9979472	.0036857	-0.56	0.578	.9907495	1.005197
	pyperstudw~e	1.000868	.0125221	0.07	0.945	.9766237	1.025714
	pyperstudb~k	1.002867	.0137287	0.21	0.834	.9763166	1.030139
	pyperstudl~o	1.008407	.0134239	0.63	0.529	.9824367	1.035063
	pyperstudm~e	.9892983	.0079299	-1.34	0.180	.9738774	1.004963
	Asian	.3786783	.4250702	-0.87	0.387	.0419554	3.417848
	Black	1.616388	.3274299	2.37	0.018	1.086723	2.404211
	Latino	1.454452	.3875475	1.41	0.160	.8627634	2.451925
	Other	.7270843	.3965576	-0.58	0.559	.2496506	2.117566
	AsianMale	9.059169	14.65116	1.36	0.173	.3805945	215.6325
	BlackMale	1.229563	.4450989	0.57	0.568	.6048086	2.499676
	LatinoMale	.6530041	.2416007	-1.15	0.249	.3162172	1.348486
	OtherMale	.9271795	.7301857	-0.10	0.924	.1980676	4.340246
3							
	tenure	.9940958	.0074258	-0.79	0.428	.9796475	1.008757
	exper	1.07504	.0116096	6.70	0.000	1.052525	1.098037
	male	1.557095	.1960861	3.52	0.000	1.21653	1.993001
	age	.9923125	.00949	-0.81	0.420	.9738856	1.011088
	pyacctrati~2	1.618096	.1553186	5.01	0.000	1.340599	1.953034
	pyperstudlep	1.016342	.0080286	2.05	0.040	1.000727	1.0322
	pyperstude~s	1.020285	.0044385	4.62	0.000	1.011623	1.029022
	pyperstudw~e	1.011158	.0141104	0.80	0.427	.983877	1.039196
	pyperstudb~k	.9998032	.0154099	-0.01	0.990	.9700519	1.030467
	pyperstudl~o	.9930203	.0149826	-0.46	0.642	.9640848	1.022824
	pyperstudm~e	.9821775	.0091425	-1.93	0.053	.964421	1.000261
	Asian	.4855507	.5640074	-0.62	0.534	.0498303	4.731249
	Black	1.363797	.3396803	1.25	0.213	.8370308	2.222072
	Latino	2.033639	.5749757	2.51	0.012	1.168454	3.539453
	Other	.6020567	.3674262	-0.83	0.406	.1820379	1.991191
	AsianMale	3.094617	5.766999	0.61	0.544	.0802312	119.3632
	BlackMale	1.911368	.7836383	1.58	0.114	.8557789	4.269007
	LatinoMale	.4216432	.1757699	-2.07	0.038	.186254	.9545188
	OtherMale	1.112804	.9949239	0.12	0.905	.1929236	6.418772

-----> region = 12

```

Iteration 0:  log likelihood = -892.86367
Iteration 1:  log likelihood = -829.72769
Iteration 2:  log likelihood = -827.66131
Iteration 3:  log likelihood = -827.40728
Iteration 4:  log likelihood = -827.36394
Iteration 5:  log likelihood = -827.3543

```

```

Iteration 6:  log likelihood = -827.35196
Iteration 7:  log likelihood = -827.35147
Iteration 8:  log likelihood = -827.35137
Iteration 9:  log likelihood = -827.35135
Iteration 10: log likelihood = -827.35134

```

```

Multinomial logistic regression      Number of obs   =      827
                                     LR chi2(38)        =     131.02
                                     Prob > chi2         =     0.0000
Log likelihood = -827.35134          Pseudo R2        =     0.0734

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.00776	.0125769	0.62	0.536	.9834088	1.032714
	exper	1.002453	.0168681	0.15	0.884	.9699312	1.036065
	male	.8076013	.1714864	-1.01	0.314	.532663	1.22445
	age	.9707723	.0148603	-1.94	0.053	.9420791	1.000339
pyacctrati~2		1.03807	.1788374	0.22	0.828	.7405985	1.455026
pyperstudlep		.9886613	.0196936	-0.57	0.567	.9508063	1.028023
pyperstude~s		.9924046	.0079654	-0.95	0.342	.9769149	1.00814
pyperstudw~e		.8850847	.0509378	-2.12	0.034	.7906732	.9907696
pyperstudb~k		.9085305	.0560594	-1.55	0.120	.80504	1.025325
pyperstudl~o		.9101093	.0527137	-1.63	0.104	.8124409	1.019519
pyperstudm~e		1.005015	.0101067	0.50	0.619	.9853998	1.02502
	Asian	2.98e-07	.0004678	-0.01	0.992	0	.
	Black	1.041727	.3644897	0.12	0.907	.5247196	2.068144
	Latino	1.109538	.6100269	0.19	0.850	.3777017	3.25938
	Other	.1599658	.1851656	-1.58	0.113	.0165476	1.546395
	AsianMale	3383845	5.32e+09	0.01	0.992	0	.
	BlackMale	.8292369	.4524799	-0.34	0.731	.284586	2.41626
	LatinoMale	3.187255	3.162898	1.17	0.243	.4557431	22.29017
	OtherMale	17.36958	32.2177	1.54	0.124	.4580906	658.6085
3							
	tenure	1.003824	.0110068	0.35	0.728	.9824813	1.025631
	exper	1.027121	.0162541	1.69	0.091	.9957523	1.059477
	male	1.50868	.292809	2.12	0.034	1.03132	2.206992
	age	.9920696	.0147836	-0.53	0.593	.9635134	1.021472
pyacctrati~2		1.637167	.2381385	3.39	0.001	1.23106	2.177242
pyperstudlep		1.033832	.019615	1.75	0.079	.9960938	1.073001
pyperstude~s		1.016577	.0078904	2.12	0.034	1.00123	1.032161
pyperstudw~e		.9455299	.0556784	-0.95	0.342	.8424641	1.061204
pyperstudb~k		.9408572	.0590146	-0.97	0.331	.8320179	1.063934
pyperstudl~o		.9282528	.0553935	-1.25	0.212	.8257921	1.043426
pyperstudm~e		.9847273	.0117667	-1.29	0.198	.961933	1.008062
	Asian	.8181408	1.029876	-0.16	0.873	.0693961	9.645422
	Black	1.654894	.6312971	1.32	0.187	.7835376	3.495268
	Latino	.9956467	.6210772	-0.01	0.994	.2931784	3.38126
	Other	2.58e-07	.0003204	-0.01	0.990	0	.
	AsianMale	6.32e-07	.0012501	-0.01	0.994	0	.
	BlackMale	.5523606	.3112549	-1.05	0.292	.1830506	1.666765
	LatinoMale	2.972426	3.209278	1.01	0.313	.3581651	24.66828
	OtherMale	7499267	9.31e+09	0.01	0.990	0	.

```

-----> region = 13

```

```

Iteration 0:  log likelihood = -1544.5731
Iteration 1:  log likelihood = -1472.1019
Iteration 2:  log likelihood = -1471.1306
Iteration 3:  log likelihood = -1471.1025
Iteration 4:  log likelihood = -1471.0979

```

```

Iteration 5:  log likelihood = -1471.0968
Iteration 6:  log likelihood = -1471.0966
Iteration 7:  log likelihood = -1471.0965
Iteration 8:  log likelihood = -1471.0965

```

```

Multinomial logistic regression          Number of obs   =      1429
                                         LR chi2(38)      =      146.95
                                         Prob > chi2      =      0.0000
Log likelihood = -1471.0965             Pseudo R2       =      0.0476

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005998	.0106592	0.56	0.573	.9853213	1.027108
	exper	.9813221	.0112713	-1.64	0.101	.9594776	1.003664
	male	.9700502	.1538184	-0.19	0.848	.7109209	1.323632
	age	.9841263	.0096617	-1.63	0.103	.9653708	1.003246
pyacctrati~2		1.091784	.1310759	0.73	0.465	.8628682	1.381431
pyperstudlep		.9983104	.0073362	-0.23	0.818	.9840348	1.012793
pyperstude~s		1.006128	.0068551	0.90	0.370	.9927813	1.019654
pyperstudw~e		.9708527	.0186292	-1.54	0.123	.9350183	1.008061
pyperstudb~k		.9669368	.0218485	-1.49	0.137	.9250489	1.010721
pyperstudl~o		.9676216	.0197378	-1.61	0.107	.9296993	1.007091
pyperstudm~e		1.00472	.0078243	0.60	0.545	.9895009	1.020173
	Asian	1.413904	1.17407	0.42	0.677	.2777179	7.198398
	Black	2.172943	.611006	2.76	0.006	1.252277	3.770476
	Latino	1.450676	.317333	1.70	0.089	.9448683	2.227252
	Other	1.149681	1.16368	0.14	0.890	.158129	8.358793
	AsianMale	538032.2	4.53e+08	0.02	0.987	0	.
	BlackMale	1.39128	.6790455	0.68	0.499	.534523	3.621284
	LatinoMale	2.157711	.8035469	2.07	0.039	1.039919	4.476996
	OtherMale	1.654012	2.024333	0.41	0.681	.150231	18.21033
3							
	tenure	1.018772	.010328	1.83	0.067	.998729	1.039217
	exper	1.024172	.0127608	1.92	0.055	.9994646	1.049491
	male	.9013297	.1504667	-0.62	0.534	.6498083	1.250207
	age	.994134	.0111198	-0.53	0.599	.9725768	1.016169
pyacctrati~2		1.529117	.1900206	3.42	0.001	1.19857	1.950823
pyperstudlep		1.000272	.0078116	0.03	0.972	.9850779	1.0157
pyperstude~s		1.028056	.0074914	3.80	0.000	1.013478	1.042844
pyperstudw~e		.9670055	.0199117	-1.63	0.103	.9287563	1.00683
pyperstudb~k		.9515276	.0235712	-2.01	0.045	.9064326	.9988661
pyperstudl~o		.9508699	.0207366	-2.31	0.021	.9110833	.992394
pyperstudm~e		.9859787	.0103734	-1.34	0.180	.9658554	1.006521
	Asian	.5530512	.6480975	-0.51	0.613	.055626	5.498608
	Black	.6792042	.2629716	-1.00	0.318	.3180062	1.450658
	Latino	1.220517	.2888723	0.84	0.400	.7675053	1.940915
	Other	3.49861	3.106127	1.41	0.158	.6140222	19.93458
	AsianMale	5.992058	8123.96	0.00	0.999	0	.
	BlackMale	3.483002	2.071821	2.10	0.036	1.085493	11.17585
	LatinoMale	1.757176	.7318799	1.35	0.176	.7767512	3.975104
	OtherMale	.1783586	.2564296	-1.20	0.230	.0106538	2.985954

```

-----> region = 14

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0:  log likelihood = -242.51841
Iteration 1:  log likelihood = -206.14306
Iteration 2:  log likelihood = -202.74889
Iteration 3:  log likelihood = -199.54376

```

```

Iteration 4: log likelihood = -199.43552
Iteration 5: log likelihood = -199.41193
Iteration 6: log likelihood = -199.40634
Iteration 7: log likelihood = -199.40513
Iteration 8: log likelihood = -199.40487
Iteration 9: log likelihood = -199.40482
Iteration 10: log likelihood = -199.4048

```

```

Multinomial logistic regression
Number of obs   =      254
LR chi2(34)     =      86.23
Prob > chi2     =      0.0000
Pseudo R2      =      0.1778

Log likelihood = -199.4048

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.000933	.0349327	0.03	0.979	.9347555	1.071796
	exper	1.072016	.0592027	1.26	0.208	.9620399	1.194564
	male	6.134415	3.81769	2.91	0.004	1.811493	20.7735
	age	.9077183	.0472187	-1.86	0.063	.8197327	1.005148
pyacctrati~2		.7513496	.3558291	-0.60	0.546	.2969781	1.900902
pyperstudlep		1.04131	.1315427	0.32	0.749	.8129286	1.333851
pyperstude~s		.9737255	.0242836	-1.07	0.286	.927275	1.022503
pyperstudw~e		.6449947	.2360129	-1.20	0.231	.3148394	1.321366
pyperstudb~k		.636041	.2557745	-1.13	0.260	.2891925	1.398889
pyperstudl~o		.6614364	.2386772	-1.15	0.252	.3260879	1.341657
pyperstudm~e		1.009268	.0439352	0.21	0.832	.9267274	1.099159
	Asian	(omitted)					
	Black	2.05e-06	.0025832	-0.01	0.992	0	.
	Latino	5.75e-06	.006056	-0.01	0.991	0	.
	Other	6.34e-06	.0098083	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	569407.7	2.27e+09	0.00	0.997	0	.
	LatinoMale	2180495	2.30e+09	0.01	0.989	0	.
	OtherMale	1.93e+14	3.85e+18	0.00	0.999	0	.
3							
	tenure	1.006833	.0223898	0.31	0.759	.9638926	1.051687
	exper	1.048284	.0321736	1.54	0.124	.987084	1.113278
	male	3.826571	1.341226	3.83	0.000	1.925133	7.606042
	age	.9414594	.0253736	-2.24	0.025	.8930188	.9925276
pyacctrati~2		2.22423	.5348638	3.32	0.001	1.388322	3.563439
pyperstudlep		1.046221	.058999	0.80	0.423	.9367466	1.16849
pyperstude~s		1.034747	.0146778	2.41	0.016	1.006376	1.063919
pyperstudw~e		1.020142	.2339083	0.09	0.931	.6508612	1.598941
pyperstudb~k		.992248	.2482792	-0.03	0.975	.607622	1.620343
pyperstudl~o		1.014375	.2299532	0.06	0.950	.6504824	1.581835
pyperstudm~e		.9505733	.0258672	-1.86	0.062	.9012029	1.002648
	Asian	(omitted)					
	Black	2.846937	3.813011	0.78	0.435	.2062203	39.30289
	Latino	2.679947	2.792488	0.95	0.344	.3476797	20.65727
	Other	1.677188	2.671677	0.32	0.745	.0739026	38.06305
	AsianMale	(omitted)					
	BlackMale	3766510	6.03e+09	0.01	0.992	0	.
	LatinoMale	1.19729	1.964557	0.11	0.913	.0480305	29.8457
	OtherMale	.0749499	2274.704	-0.00	1.000	0	.

```
-----> region = 15
```

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note: Asian omitted because of collinearity
note: Other omitted because of collinearity
note: AsianMale omitted because of collinearity

```

note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -272.36452  
 Iteration 1: log likelihood = -239.87152  
 Iteration 2: log likelihood = -236.45057  
 Iteration 3: log likelihood = -236.3656  
 Iteration 4: log likelihood = -236.3634  
 Iteration 5: log likelihood = -236.36301  
 Iteration 6: log likelihood = -236.36292  
 Iteration 7: log likelihood = -236.3629

Multinomial logistic regression

Number of obs = 268  
 LR chi2(30) = 72.00  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1322

Log likelihood = -236.3629

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.067301	.0383137	1.81	0.070	.9947879	1.145099
	exper	.9405068	.037644	-1.53	0.125	.8695457	1.017259
	male	1.415896	.7850288	0.63	0.531	.4776303	4.197311
	age	1.050048	.0329	1.56	0.119	.9875054	1.116552
pyacctrati~2		.3055756	.1613461	-2.25	0.025	.1085621	.8601203
pyperstudlep		.9397897	.043726	-1.33	0.182	.8578798	1.02952
pyperstude~s		.9979793	.0170623	-0.12	0.906	.965092	1.031987
pyperstudw~e		1.032945	.326329	0.10	0.918	.5561156	1.918623
pyperstudb~k		1.218756	.4273175	0.56	0.573	.6130144	2.423055
pyperstudl~o		1.029032	.3251922	0.09	0.928	.5539039	1.911717
pyperstudm~e		.9662631	.0387168	-0.86	0.392	.8932828	1.045206
	Asian	(omitted)					
	Black	.7904674	5.40e+09	0.02	0.981	0	.
	Latino	7.245263	4.985106	2.88	0.004	1.880982	27.90768
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	7.30e-07	.0004987	-0.02	0.983	0	.
	LatinoMale	.6034704	.5735371	-0.53	0.595	.0936862	3.887194
	OtherMale	(omitted)					
3							
	tenure	.980802	.0208481	-0.91	0.362	.9407799	1.022527
	exper	1.069114	.0302797	2.36	0.018	1.011384	1.130139
	male	.9589042	.3260076	-0.12	0.902	.4924751	1.867094
	age	.9881032	.0249742	-0.47	0.636	.9403473	1.038284
pyacctrati~2		1.1295	.306538	0.45	0.654	.6635542	1.922631
pyperstudlep		1.023346	.0282801	0.84	0.404	.9693928	1.080303
pyperstude~s		1.010462	.0112064	0.94	0.348	.9887351	1.032667
pyperstudw~e		.9588187	.2038587	-0.20	0.843	.6320606	1.454502
pyperstudb~k		.885667	.2015432	-0.53	0.594	.5669833	1.383473
pyperstudl~o		.9480615	.20113	-0.25	0.801	.6255402	1.436871
pyperstudm~e		1.013842	.0218227	0.64	0.523	.9719595	1.057528
	Asian	(omitted)					
	Black	1.271444	1309.483	0.00	1.000	0	.
	Latino	.9874384	.569976	-0.02	0.983	.318547	3.060881
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	3.503539	3608.36	0.00	0.999	0	.
	LatinoMale	.7123028	.6071109	-0.40	0.691	.1340193	3.785838
	OtherMale	(omitted)					

-----> region = 16

note: Asian omitted because of collinearity

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -506.93704  
 Iteration 1: log likelihood = -455.05451  
 Iteration 2: log likelihood = -452.57734  
 Iteration 3: log likelihood = -452.43233  
 Iteration 4: log likelihood = -452.39897  
 Iteration 5: log likelihood = -452.39213  
 Iteration 6: log likelihood = -452.39069  
 Iteration 7: log likelihood = -452.39037  
 Iteration 8: log likelihood = -452.39029  
 Iteration 9: log likelihood = -452.39027

Multinomial logistic regression	Number of obs	=	468
	LR chi2(34)	=	109.09
	Prob > chi2	=	0.0000
Log likelihood = -452.39027	Pseudo R2	=	0.1076

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.072095	.0242197	3.08	0.002	1.02566	1.120631
	exper	.9209834	.0241517	-3.14	0.002	.8748429	.9695575
	male	1.208714	.3290602	0.70	0.486	.7089128	2.060889
	age	.9681724	.0202154	-1.55	0.121	.9293508	1.008616
pyacctrati~2		.8554419	.1905653	-0.70	0.483	.5528023	1.323766
pyperstudlep		.9990766	.0187029	-0.05	0.961	.9630839	1.036414
pyperstude~s		.9792978	.0089	-2.30	0.021	.9620086	.9968977
pyperstudw~e		.9397032	.0312134	-1.87	0.061	.8804749	1.002916
pyperstudb~k		.981422	.0363056	-0.51	0.612	.9127828	1.055223
pyperstudl~o		.9573319	.0300646	-1.39	0.165	.9001832	1.018109
pyperstudm~e		1.002705	.012829	0.21	0.833	.9778733	1.028167
	Asian	(omitted)					
	Black	.3320182	.4260711	-0.86	0.390	.0268433	4.106659
	Latino	3.629771	2.431583	1.92	0.054	.9764711	13.49271
	Other	5.586828	9.44493	1.02	0.309	.2033053	153.526
	AsianMale	(omitted)					
	BlackMale	1.31e-06	.0011012	-0.02	0.987	0	.
	LatinoMale	.7187118	1.010743	-0.23	0.814	.0456549	11.31415
	OtherMale	1.21e+07	3.63e+10	0.01	0.996	0	.
3							
	tenure	1.027296	.0165251	1.67	0.094	.9954123	1.0602
	exper	1.028624	.0224837	1.29	0.197	.9854872	1.073649
	male	2.217751	.5457652	3.24	0.001	1.369118	3.592399
	age	.9560888	.0198619	-2.16	0.031	.917942	.9958208
pyacctrati~2		1.546136	.2919653	2.31	0.021	1.067851	2.238642
pyperstudlep		1.037345	.0169036	2.25	0.024	1.004738	1.07101
pyperstude~s		1.005103	.0077157	0.66	0.507	.9900937	1.02034
pyperstudw~e		1.031723	.0416265	0.77	0.439	.9532788	1.116622
pyperstudb~k		1.04494	.0453026	1.01	0.311	.9598166	1.137613
pyperstudl~o		1.021488	.0400154	0.54	0.587	.9459943	1.103006
pyperstudm~e		.9648013	.0164232	-2.11	0.035	.9331434	.9975332
	Asian	(omitted)					
	Black	1.507654	1.543678	0.40	0.688	.202657	11.2161
	Latino	2.03389	1.454095	0.99	0.321	.500928	8.258089
	Other	1.77e-06	.0016074	-0.01	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	.2342922	.3750593	-0.91	0.365	.0101657	5.399792
	LatinoMale	1.684273	2.27173	0.39	0.699	.1197625	23.68669
	OtherMale	387124.3	1.74e+09	0.00	0.998	0	.

-----> region = 17

note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -460.40979  
Iteration 1: log likelihood = -409.50513  
Iteration 2: log likelihood = -406.88023  
Iteration 3: log likelihood = -406.70356  
Iteration 4: log likelihood = -406.66383  
Iteration 5: log likelihood = -406.65417  
Iteration 6: log likelihood = -406.65213  
Iteration 7: log likelihood = -406.6517  
Iteration 8: log likelihood = -406.65161  
Iteration 9: log likelihood = -406.65158

Multinomial logistic regression                      Number of obs    =        452  
   LR chi2(34)        =        107.52  
   Prob > chi2        =        0.0000  
Log likelihood = -406.65158                               Pseudo R2        =        0.1168

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9975173	.0227854	-0.11	0.913	.9538436	1.043191
	exper	1.026046	.031248	0.84	0.399	.966593	1.089156
	male	1.92147	.6401862	1.96	0.050	1.000079	3.691755
	age	.9704097	.0266368	-1.09	0.274	.9195821	1.024047
pyacctrati~2		1.078046	.3139725	0.26	0.796	.6091604	1.907845
pyperstudlep		.9864094	.0431533	-0.31	0.754	.905355	1.07472
pyperstude~s		.9812933	.0133543	-1.39	0.165	.9554654	1.007819
pyperstudw~e		1.120597	.1180283	1.08	0.280	.9115821	1.377538
pyperstudb~k		1.136153	.1232467	1.18	0.239	.9185455	1.405312
pyperstudl~o		1.120477	.1199065	1.06	0.288	.9084745	1.381953
pyperstudm~e		1.005131	.0142039	0.36	0.717	.9776735	1.033359
	Asian	7.47e-07	.0015812	-0.01	0.995	0	.
	Black	2.14e-06	.001992	-0.01	0.989	0	.
	Latino	2.783172	1.890805	1.51	0.132	.7349515	10.53953
	Other	.8907752	2476.09	-0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	1.04e+12	1.27e+15	0.02	0.982	0	.
	LatinoMale	2.424625	2.408488	0.89	0.373	.3460256	16.98952
	OtherMale	(omitted)					
3							
	tenure	.9974686	.017027	-0.15	0.882	.9646485	1.031405
	exper	1.047848	.0232372	2.11	0.035	1.003279	1.094396
	male	3.519892	.9223525	4.80	0.000	2.106107	5.88272
	age	.9839524	.0197159	-0.81	0.419	.946059	1.023364
pyacctrati~2		1.490615	.3258616	1.83	0.068	.9711481	2.287946
pyperstudlep		.9889181	.0254503	-0.43	0.665	.9402735	1.040079
pyperstude~s		1.041583	.0116389	3.65	0.000	1.019019	1.064647
pyperstudw~e		1.053405	.0852173	0.64	0.520	.8989506	1.234398
pyperstudb~k		1.019536	.0860699	0.23	0.819	.8640589	1.202988
pyperstudl~o		1.031389	.0850896	0.37	0.708	.8774014	1.212402
pyperstudm~e		.9717532	.0148296	-1.88	0.060	.9431181	1.001258
	Asian	4.19e-06	.0056911	-0.01	0.993	0	.
	Black	2.084427	2.867218	0.53	0.593	.1406446	30.89231
	Latino	2.155143	1.229395	1.35	0.178	.7045477	6.592372
	Other	3.83e+07	5.43e+10	0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	806665.1	6.37e+08	0.02	0.986	0	.
	LatinoMale	1.344935	1.201963	0.33	0.740	.2333383	7.752055
	OtherMale	(omitted)					



-----> region = 18

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -406.99158  
 Iteration 1: log likelihood = -358.6355  
 Iteration 2: log likelihood = -355.84989  
 Iteration 3: log likelihood = -355.70264  
 Iteration 4: log likelihood = -355.67088  
 Iteration 5: log likelihood = -355.66301  
 Iteration 6: log likelihood = -355.66142  
 Iteration 7: log likelihood = -355.66107  
 Iteration 8: log likelihood = -355.661  
 Iteration 9: log likelihood = -355.66098

Multinomial logistic regression	Number of obs	=	387
	LR chi2(34)	=	102.66
	Prob > chi2	=	0.0000
Log likelihood = -355.66098	Pseudo R2	=	0.1261

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9995572	.0239144	-0.02	0.985	.9537678	1.047545
	exper	1.002641	.0294181	0.09	0.928	.9466087	1.061989
	male	1.237548	.4460167	0.59	0.554	.6106408	2.508061
	age	.9633488	.0227085	-1.58	0.113	.9198535	1.008901
pyacctrati~2		.3141254	.1139671	-3.19	0.001	.1542698	.6396249
pyperstudlep		.9743774	.0243658	-1.04	0.299	.9277727	1.023323
pyperstude~s		1.00657	.0161363	0.41	0.683	.9754355	1.038699
pyperstudw~e		.5979242	.0891465	-3.45	0.001	.4464139	.8008563
pyperstudb~k		.6111462	.0932419	-3.23	0.001	.4531878	.824161
pyperstudl~o		.5970721	.089509	-3.44	0.001	.4450619	.8010012
pyperstudm~e		.9752926	.0171745	-1.42	0.155	.9422055	1.009542
	Asian	(omitted)					
	Black	12.92706	15.19167	2.18	0.029	1.29176	129.3652
	Latino	2.535341	1.237294	1.91	0.057	.9741679	6.598403
	Other	4.274799	11766.8	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	161904	8.59e+07	0.02	0.982	0	.
	LatinoMale	.7884144	.5069673	-0.37	0.712	.2235701	2.780323
	OtherMale	4768502	1.80e+10	0.00	0.997	0	.
3							
	tenure	.9955651	.0187377	-0.24	0.813	.9595089	1.032976
	exper	1.059403	.0275948	2.22	0.027	1.006676	1.114892
	male	1.38137	.425223	1.05	0.294	.7555919	2.525416
	age	.965055	.0220076	-1.56	0.119	.9228707	1.009168
pyacctrati~2		1.475062	.3862374	1.48	0.138	.8829324	2.464297
pyperstudlep		.984829	.0194024	-0.78	0.438	.9475258	1.023601
pyperstude~s		1.062381	.015599	4.12	0.000	1.032243	1.093399
pyperstudw~e		.7983287	.1144782	-1.57	0.116	.6027286	1.057406
pyperstudb~k		.7632318	.1146917	-1.80	0.072	.5685196	1.024631
pyperstudl~o		.7690517	.111406	-1.81	0.070	.5789608	1.021555
pyperstudm~e		.9581637	.0207925	-1.97	0.049	.9182656	.9997954
	Asian	(omitted)					
	Black	3.996749	5.037109	1.10	0.272	.3380154	47.25822
	Latino	1.138081	.4734625	0.31	0.756	.5035673	2.572104
	Other	3157688	4.92e+09	0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	237904.2	1.26e+08	0.02	0.981	0	.
	LatinoMale	.7907624	.4639771	-0.40	0.689	.250384	2.497385
	OtherMale	3.98e-07	.0017585	-0.00	0.997	0	.

-----> region = 19

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -791.08657  
 Iteration 1: log likelihood = -723.85766  
 Iteration 2: log likelihood = -720.97851  
 Iteration 3: log likelihood = -720.85858  
 Iteration 4: log likelihood = -720.83068  
 Iteration 5: log likelihood = -720.82471  
 Iteration 6: log likelihood = -720.82347  
 Iteration 7: log likelihood = -720.82319  
 Iteration 8: log likelihood = -720.82312  
 Iteration 9: log likelihood = -720.82311

Multinomial logistic regression	Number of obs	=	729
	LR chi2(36)	=	140.53
	Prob > chi2	=	0.0000
Log likelihood = -720.82311	Pseudo R2	=	0.0888

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.023493	.0147212	1.61	0.106	.9950427 1.052756
exper	1.019838	.01791	1.12	0.263	.9853323 1.055552
male	1.241385	.3978368	0.67	0.500	.6623922 2.326473
age	.9722642	.0142673	-1.92	0.055	.9446991 1.000634
pyacctrati~2	.901919	.1853659	-0.50	0.615	.6028722 1.349304
pyperstudlep	1.014472	.0063865	2.28	0.022	1.002031 1.027066
pyperstude~s	.9847381	.0100715	-1.50	0.133	.9651949 1.004677
pyperstudw~e	.8313313	.0926701	-1.66	0.097	.6681732 1.03433
pyperstudb~k	.8793287	.1029117	-1.10	0.272	.6990872 1.106041
pyperstudl~o	.859762	.0903012	-1.44	0.150	.6998036 1.056283
pyperstudm~e	1.005981	.0102346	0.59	0.558	.9861205 1.026242
Asian	.3745243	1437.106	-0.00	1.000	0 .
Black	.7811449	.5189007	-0.37	0.710	.2124679 2.871903
Latino	1.11753	.2832463	0.44	0.661	.6800104 1.836549
Other	3257883	4.04e+09	0.01	0.990	0 .
AsianMale	(omitted)				
BlackMale	3.822199	4.204145	1.22	0.223	.4426422 33.00455
LatinoMale	1.09221	.4257067	0.23	0.821	.5087872 2.344638
OtherMale	2.11e-07	.0002616	-0.01	0.990	0 .
3					
tenure	1.049148	.0159213	3.16	0.002	1.018403 1.080822
exper	1.077349	.0219627	3.65	0.000	1.035152 1.121267
male	1.712689	.5910048	1.56	0.119	.8708653 3.368265
age	.9794787	.0175354	-1.16	0.247	.945706 1.014458
pyacctrati~2	1.229291	.2797892	0.91	0.364	.7868994 1.920392
pyperstudlep	1.024627	.0073064	3.41	0.001	1.010406 1.039047
pyperstude~s	1.0242	.0124463	1.97	0.049	1.000094 1.048887
pyperstudw~e	.9016428	.1083308	-0.86	0.389	.712466 1.141051
pyperstudb~k	.9097121	.1206849	-0.71	0.476	.7014251 1.17985
pyperstudl~o	.880177	.1025164	-1.10	0.273	.7005328 1.105889
pyperstudm~e	.9707482	.0154706	-1.86	0.062	.9408951 1.001549
Asian	477089.2	1.29e+09	0.00	0.996	0 .
Black	.7100307	.5538487	-0.44	0.661	.1539254 3.275247
Latino	.9144782	.2624652	-0.31	0.755	.5210378 1.605009
Other	1.119174	2003.965	0.00	1.000	0 .
AsianMale	(omitted)				
BlackMale	3.342502	4.215113	0.96	0.339	.2822611 39.58151
LatinoMale	.5633558	.2479503	-1.30	0.192	.2377621 1.33482

```

OtherMale | 1.89e-06 .003724 -0.01 0.995 0 .
-----

```

```

-----> region = 20

```

```

Iteration 0: log likelihood = -2090.3193
Iteration 1: log likelihood = -1981.5148
Iteration 2: log likelihood = -1976.8499
Iteration 3: log likelihood = -1976.5311
Iteration 4: log likelihood = -1976.4776
Iteration 5: log likelihood = -1976.4657
Iteration 6: log likelihood = -1976.4627
Iteration 7: log likelihood = -1976.4621
Iteration 8: log likelihood = -1976.462
Iteration 9: log likelihood = -1976.4619
Iteration 10: log likelihood = -1976.4619

```

```

Multinomial logistic regression      Number of obs   =      2048
                                     LR chi2(38)       =      227.71
                                     Prob > chi2       =      0.0000
Log likelihood = -1976.4619          Pseudo R2       =      0.0545

```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.997434	.0083944	-0.31	0.760	.9811162	1.014023
exper	1.010798	.0105222	1.03	0.302	.9903839	1.031633
male	1.466052	.2191949	2.56	0.011	1.093662	1.965241
age	.9708009	.0085217	-3.38	0.001	.9542414	.9876477
pyacctrati~2	1.037245	.1130875	0.34	0.737	.8376786	1.284355
pyperstudlep	.9799815	.0072645	-2.73	0.006	.9658464	.9943235
pyperstude~s	1.004774	.005284	0.91	0.365	.9944706	1.015184
pyperstudw~e	.9513493	.0388921	-1.22	0.222	.878096	1.030714
pyperstudb~k	.9468824	.0410008	-1.26	0.207	.8698378	1.030751
pyperstudl~o	.9523996	.0394256	-1.18	0.239	.8781786	1.032894
pyperstudm~e	.9993946	.0075021	-0.08	0.936	.9847983	1.014207
Asian	.5242816	.592047	-0.57	0.567	.0573247	4.794985
Black	1.375719	.3760868	1.17	0.243	.8050687	2.35086
Latino	.9902755	.1542598	-0.06	0.950	.729727	1.343853
Other	.623672	.506825	-0.58	0.561	.1268324	3.066778
AsianMale	3.67e+07	1.46e+11	0.00	0.997	0	.
BlackMale	.9336018	.4118748	-0.16	0.876	.3932233	2.216584
LatinoMale	.991615	.2352689	-0.04	0.972	.6228571	1.578693
OtherMale	.5868847	.815932	-0.38	0.701	.0384722	8.952787
3						
tenure	1.01889	.0084411	2.26	0.024	1.002479	1.035569
exper	1.082116	.0139568	6.12	0.000	1.055104	1.109819
male	1.62611	.2712619	2.91	0.004	1.172616	2.254989
age	.9543312	.0111654	-4.00	0.000	.9326964	.9764679
pyacctrati~2	1.263814	.1512548	1.96	0.050	.999563	1.597923
pyperstudlep	1.014531	.0069381	2.11	0.035	1.001023	1.028221
pyperstude~s	1.033112	.0062472	5.39	0.000	1.02094	1.045429
pyperstudw~e	1.009955	.0474109	0.21	0.833	.9211775	1.107287
pyperstudb~k	.9770068	.0487862	-0.47	0.641	.8859176	1.077462
pyperstudl~o	.9805379	.0468189	-0.41	0.681	.8929375	1.076732
pyperstudm~e	.9827139	.0088865	-1.93	0.054	.9654502	1.000286
Asian	4.42e-07	.0005751	-0.01	0.991	0	.
Black	1.337494	.4285343	0.91	0.364	.7137833	2.506211
Latino	.8523271	.145463	-0.94	0.349	.6100078	1.190905
Other	4.35e-07	.000439	-0.01	0.988	0	.
AsianMale	1523234	1.15e+10	0.00	0.998	0	.
BlackMale	.8554271	.4577413	-0.29	0.770	.2997111	2.441536

```

LatinoMale | .8709094 .2329358 -0.52 0.605 .5155945 1.471085
OtherMale | 3841259 3.88e+09 0.02 0.988 0 .
-----
by pyschooltype:
. bysort pyschooltype : mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2005a==1, rrr

-----> pyschooltype =
no observations

-----> pyschooltype = B

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -350.7425
Iteration 1: log likelihood = -327.55515
Iteration 2: log likelihood = -322.14884
Iteration 3: log likelihood = -321.9914
Iteration 4: log likelihood = -321.97238
Iteration 5: log likelihood = -321.96917
Iteration 6: log likelihood = -321.96843
Iteration 7: log likelihood = -321.96825
Iteration 8: log likelihood = -321.96822
Iteration 9: log likelihood = -321.96821

Multinomial logistic regression
Log likelihood = -321.96821
Number of obs = 387
LR chi2(32) = 57.55
Prob > chi2 = 0.0037
Pseudo R2 = 0.0820

-----
admin | RRR Std. Err. z P>|z| [95% Conf. Interval]
-----+-----
1 | (base outcome)
-----+-----
2 |
tenure | 1.049173 .0306662 1.64 0.101 .9907574 1.111033
exper | 1.0213 .0482429 0.45 0.655 .9309911 1.12037
male | 1.284791 .5976066 0.54 0.590 .5163038 3.197125
age | .9324639 .0371087 -1.76 0.079 .8624963 1.008107
pyacctrati~2 | 1.121177 .4396629 0.29 0.771 .519853 2.418064
pyperstudlep | .986637 .0328384 -0.40 0.686 .9243293 1.053145
pyperstude~s | .9680531 .0150068 -2.09 0.036 .9390827 .9979173
pyperstudw~e | .9959537 .0346562 -0.12 0.907 .9302933 1.066248
pyperstudb~k | 1.056929 .0433222 1.35 0.177 .9753399 1.145343
pyperstudl~o | 1.029903 .0381197 0.80 0.426 .9578355 1.107393
pyperstudm~e | .9607096 .0222058 -1.73 0.083 .9181582 1.005233
Asian | (omitted)
Black | .8585905 .7058452 -0.19 0.853 .1714018 4.300875
Latino | 1.12029 1.386328 0.09 0.927 .0990812 12.66688
Other | 6.57e-07 .0005179 -0.02 0.986 0 .
AsianMale | (omitted)
BlackMale | 1.657899 1.676415 0.50 0.617 .2284809 12.03001
LatinoMale | 1.611841 2.346976 0.33 0.743 .0928756 27.97325
OtherMale | (omitted)
-----
3 |
tenure | .9669056 .015994 -2.03 0.042 .9360607 .998767
exper | 1.015055 .0196523 0.77 0.440 .9772587 1.054313
male | 1.25129 .3180997 0.88 0.378 .7602694 2.059435
age | .9864902 .0180695 -0.74 0.458 .9517029 1.022549
pyacctrati~2 | .9029194 .1977773 -0.47 0.641 .5877601 1.387068

```

pyperstudlep		.9865712	.0184461	-0.72	0.470	.9510719	1.023395
pyperstude~s		.9971577	.0080985	-0.35	0.726	.9814107	1.013157
pyperstudw~e		.9984531	.0279418	-0.06	0.956	.945163	1.054748
pyperstudb~k		1.010635	.0313288	0.34	0.733	.9510602	1.073942
pyperstudl~o		1.0069	.0288362	0.24	0.810	.9519388	1.065034
pyperstudm~e		.9999809	.0091981	-0.00	0.998	.9821145	1.018172
Asian		(omitted)					
Black		.308373	.266157	-1.36	0.173	.0568072	1.673975
Latino		.5516255	.4845061	-0.68	0.498	.098631	3.085143
Other		.9876339	1.08229	-0.01	0.991	.115296	8.460141
AsianMale		(omitted)					
BlackMale		2.936095	2.99132	1.06	0.290	.3986193	21.62628
LatinoMale		2.649042	2.885691	0.89	0.371	.3132174	22.40431
OtherMale		(omitted)					

-----> pyschooltype = E

Iteration 0: log likelihood = -10893.698  
Iteration 1: log likelihood = -10281.717  
Iteration 2: log likelihood = -10275.618  
Iteration 3: log likelihood = -10275.612  
Iteration 4: log likelihood = -10275.612

Multinomial logistic regression	Number of obs	=	10073
	LR chi2(38)	=	1236.17
	Prob > chi2	=	0.0000
Log likelihood = -10275.612	Pseudo R2	=	0.0567

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	1.012605	.0043487	2.92	0.004	1.004117 1.021164
	exper	1.009334	.0055889	1.68	0.093	.9984387 1.020347
	male	1.482077	.1287728	4.53	0.000	1.250007 1.757231
	age	.9767717	.0042666	-5.38	0.000	.9684451 .9851699
pyacctrati~2		.9099555	.03864	-2.22	0.026	.8372884 .9889292
pyperstudlep		1.008606	.0019221	4.50	0.000	1.004846 1.012381
pyperstude~s		.9865362	.002091	-6.40	0.000	.9824464 .990643
pyperstudw~e		.9985351	.0055409	-0.26	0.792	.9877339 1.009454
pyperstudb~k		.9997149	.0062372	-0.05	0.964	.9875646 1.012015
pyperstudl~o		1.002302	.0058355	0.39	0.693	.9909295 1.013805
pyperstudm~e		1.011644	.0036714	3.19	0.001	1.004473 1.018865
	Asian	1.540349	.6251778	1.06	0.287	.6952551 3.41267
	Black	1.319809	.1264013	2.90	0.004	1.093929 1.592328
	Latino	1.100046	.0891301	1.18	0.239	.9385196 1.289373
	Other	.7635972	.2328011	-0.88	0.376	.4201022 1.38795
	AsianMale	.6654688	.5809662	-0.47	0.641	.1202302 3.683341
	BlackMale	.6852973	.1356137	-1.91	0.056	.4649802 1.010005
	LatinoMale	.8206711	.1232674	-1.32	0.188	.6113866 1.101596
	OtherMale	1.813431	1.01624	1.06	0.288	.6046313 5.438903
3						
	tenure	1.018992	.0036684	5.23	0.000	1.011828 1.026208
	exper	1.069595	.0055087	13.06	0.000	1.058853 1.080446
	male	2.464079	.1872546	11.87	0.000	2.123091 2.859833
	age	.9847831	.0042877	-3.52	0.000	.9764151 .9932229
pyacctrati~2		.9820937	.0389255	-0.46	0.648	.9086892 1.061428
pyperstudlep		.9975773	.0018357	-1.32	0.187	.9939859 1.001182
pyperstude~s		1.003557	.002018	1.77	0.077	.9996092 1.00752
pyperstudw~e		1.009477	.005713	1.67	0.096	.9983419 1.020737
pyperstudb~k		.9998593	.0063299	-0.02	0.982	.9875295 1.012343
pyperstudl~o		1.001925	.0059539	0.32	0.746	.9903237 1.013663

pyperstudm~e		.9983755	.0036636	-0.44	0.658	.9912208	1.005582
Asian		.8483788	.4332519	-0.32	0.747	.3118146	2.308252
Black		.9331104	.0918861	-0.70	0.482	.76933	1.131757
Latino		1.148116	.0904436	1.75	0.080	.9838564	1.339799
Other		.6261805	.195778	-1.50	0.134	.3392874	1.155663
AsianMale		1.073523	.9522686	0.08	0.936	.1886918	6.107588
BlackMale		.5305151	.1025093	-3.28	0.001	.3632649	.7747686
LatinoMale		.5878267	.0831537	-3.76	0.000	.4454907	.7756395
OtherMale		.4301692	.3056347	-1.19	0.235	.1068714	1.731478

-----> pyschooltype = M

Iteration 0: log likelihood = -6034.4814  
Iteration 1: log likelihood = -5733.9286  
Iteration 2: log likelihood = -5727.3951  
Iteration 3: log likelihood = -5727.3487  
Iteration 4: log likelihood = -5727.3487

Multinomial logistic regression	Number of obs	=	5738
	LR chi2(38)	=	614.27
	Prob > chi2	=	0.0000
Log likelihood = -5727.3487	Pseudo R2	=	0.0509

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	1.012849	.0050727	2.55	0.011	1.002955 1.022841
	exper	1.006458	.0062636	1.03	0.301	.9942563 1.01881
	male	1.608307	.1279052	5.98	0.000	1.376179 1.879591
	age	.9725338	.0048965	-5.53	0.000	.9629841 .9821782
	pyacctrati~2	.8919754	.0607311	-1.68	0.093	.780545 1.019314
	pyperstudlep	1.008246	.0039544	2.09	0.036	1.000526 1.016027
	pyperstude~s	.992661	.0026923	-2.72	0.007	.9873981 .9979519
	pyperstudw~e	.9998076	.0081516	-0.02	0.981	.9839576 1.015913
	pyperstudb~k	1.004776	.0090575	0.53	0.597	.9871793 1.022686
	pyperstudl~o	1.001475	.0086506	0.17	0.864	.9846632 1.018574
	pyperstudm~e	1.006909	.0042193	1.64	0.100	.9986737 1.015213
	Asian	.915893	.381272	-0.21	0.833	.4050445 2.071031
	Black	.9542911	.1134848	-0.39	0.694	.7558847 1.204776
	Latino	2.110739	.2514078	6.27	0.000	1.671279 2.665755
	Other	.7649229	.2921073	-0.70	0.483	.3618782 1.616862
	AsianMale	2.344936	1.939751	1.03	0.303	.4634576 11.86457
	BlackMale	1.482942	.267369	2.19	0.029	1.041491 2.111508
	LatinoMale	.7716708	.1245488	-1.61	0.108	.562402 1.058808
	OtherMale	1.032942	.5935295	0.06	0.955	.3349458 3.185497
3						
	tenure	1.006821	.0050933	1.34	0.179	.9968875 1.016853
	exper	1.071954	.0079339	9.39	0.000	1.056516 1.087618
	male	1.982824	.1741874	7.79	0.000	1.669197 2.355378
	age	.9665818	.0063195	-5.20	0.000	.9542748 .9790476
	pyacctrati~2	.9456298	.0728625	-0.73	0.468	.8130823 1.099785
	pyperstudlep	.9916983	.0051265	-1.61	0.107	.9817013 1.001797
	pyperstude~s	1.010524	.0032422	3.26	0.001	1.004189 1.016898
	pyperstudw~e	1.029295	.0112433	2.64	0.008	1.007492 1.051569
	pyperstudb~k	1.013433	.0122257	1.11	0.269	.9897519 1.03768
	pyperstudl~o	1.015017	.0117106	1.29	0.196	.992322 1.038231
	pyperstudm~e	.9928631	.0058866	-1.21	0.227	.9813923 1.004468
	Asian	.4357554	.3306483	-1.09	0.274	.098482 1.928097
	Black	.8120959	.1290392	-1.31	0.190	.5947773 1.108818
	Latino	1.695446	.2541447	3.52	0.000	1.263834 2.274457
	Other	.281286	.2094855	-1.70	0.089	.0653465 1.210804

AsianMale		1.331629	1.859825	0.21	0.838	.0862082	20.56921
BlackMale		1.839457	.4040783	2.77	0.006	1.195926	2.829276
LatinoMale		.84495	.1632277	-0.87	0.383	.578623	1.233861
OtherMale		3.334359	2.882591	1.39	0.164	.6125464	18.15038

-----> pyschooltype = S

Iteration 0: log likelihood = -6831.1206  
Iteration 1: log likelihood = -6538.6708  
Iteration 2: log likelihood = -6528.2176  
Iteration 3: log likelihood = -6527.9754  
Iteration 4: log likelihood = -6527.9221  
Iteration 5: log likelihood = -6527.9087  
Iteration 6: log likelihood = -6527.9062  
Iteration 7: log likelihood = -6527.9057  
Iteration 8: log likelihood = -6527.9056  
Iteration 9: log likelihood = -6527.9056

Multinomial logistic regression	Number of obs	=	7060
	LR chi2(38)	=	606.43
	Prob > chi2	=	0.0000
Log likelihood = -6527.9056	Pseudo R2	=	0.0444

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	1.004793	.0039239	1.22	0.221	.9971317 1.012513
	exper	1.007666	.005175	1.49	0.137	.9975741 1.01786
	male	1.160527	.0784322	2.20	0.028	1.016549 1.324898
	age	.9692543	.0044225	-6.84	0.000	.960625 .977961
	pyacctrati~2	.8901905	.0595216	-1.74	0.082	.7808512 1.01484
	pypersstudlep	1.018558	.0044308	4.23	0.000	1.009911 1.02728
	pypersstude~s	.9881208	.0022605	-5.22	0.000	.9837002 .9925613
	pypersstudw~e	.9974384	.0064093	-0.40	0.690	.9849552 1.01008
	pypersstudb~k	1.003254	.0071357	0.46	0.648	.9893658 1.017338
	pypersstudl~o	1.003015	.006849	0.44	0.659	.9896802 1.016529
	pypersstudm~e	1.001678	.0035623	0.47	0.637	.9947207 1.008685
	Asian	.4948632	.3230068	-1.08	0.281	.1376859 1.778611
	Black	1.244244	.1463748	1.86	0.063	.988027 1.566903
	Latino	1.542904	.1785222	3.75	0.000	1.229845 1.935653
	Other	.5760823	.2529302	-1.26	0.209	.2436478 1.362092
	AsianMale	2.237602	1.892949	0.95	0.341	.4262752 11.74561
	BlackMale	2.035232	.3374267	4.29	0.000	1.470588 2.816676
	LatinoMale	1.120828	.162	0.79	0.430	.8443253 1.487881
	OtherMale	2.412609	1.353571	1.57	0.116	.8033934 7.245119
3						
	tenure	.9950575	.004654	-1.06	0.289	.9859776 1.004221
	exper	1.051236	.0074265	7.07	0.000	1.036781 1.065893
	male	1.929469	.1783078	7.11	0.000	1.609815 2.312596
	age	.9751233	.0065195	-3.77	0.000	.9624286 .9879854
	pyacctrati~2	1.163608	.0993302	1.78	0.076	.9843393 1.375526
	pypersstudlep	1.004596	.0065534	0.70	0.482	.9918338 1.017523
	pypersstude~s	1.01809	.0030231	6.04	0.000	1.012182 1.024033
	pypersstudw~e	1.039159	.0116802	3.42	0.001	1.016516 1.062305
	pypersstudb~k	1.027209	.0125096	2.20	0.027	1.002981 1.052022
	pypersstudl~o	1.022878	.0120039	1.93	0.054	.9996194 1.046678
	pypersstudm~e	.9888999	.0052412	-2.11	0.035	.9786805 .999226
	Asian	.6104833	.6406189	-0.47	0.638	.0780646 4.774123
	Black	.5787686	.1363708	-2.32	0.020	.3647059 .9184743
	Latino	1.165919	.219088	0.82	0.414	.8067128 1.685069
	Other	.5847215	.4368786	-0.72	0.473	.1351973 2.528891

AsianMale		3.50e-06	.0016532	-0.03	0.979	0	.
BlackMale		3.248584	.9003917	4.25	0.000	1.886998	5.59264
LatinoMale		.851176	.1878591	-0.73	0.465	.552274	1.31185
OtherMale		2.568633	2.239263	1.08	0.279	.465209	14.18261

## 2005-06

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
> Male LatinoMale OtherMale if validcert2006a==1, rrr
```

```
Iteration 0: log likelihood = -25141.264
Iteration 1: log likelihood = -23974.392
Iteration 2: log likelihood = -23953.485
Iteration 3: log likelihood = -23953.448
Iteration 4: log likelihood = -23953.448
```

```
Multinomial logistic regression      Number of obs   =      23586
                                     LR chi2(38)      =      2375.63
                                     Prob > chi2      =      0.0000
Log likelihood = -23953.448          Pseudo R2       =      0.0472
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.010812	.0024695	4.40	0.000	1.005984	1.015664
	exper	1.005816	.0031169	1.87	0.061	.9997256	1.011944
	male	1.20363	.0495188	4.51	0.000	1.110385	1.304706
	age	.9727807	.0024992	-10.74	0.000	.9678947	.9776914
	pyacctrati~2	1.018514	.0315998	0.59	0.554	.9584249	1.08237
	pyperstudlep	1.008658	.0013042	6.67	0.000	1.006105	1.011217
	pyperstude~s	.9911268	.0012686	-6.96	0.000	.9886434	.9936164
	pyperstudw~e	.9955861	.0034775	-1.27	0.205	.9887937	1.002425
	pyperstudb~k	1.00023	.003938	0.06	0.953	.992541	1.007978
	pyperstudl~o	.9992387	.0037207	-0.20	0.838	.9919729	1.006558
	pyperstudm~e	1.004767	.0020971	2.28	0.023	1.000665	1.008885
	Asian	.9926128	.2558701	-0.03	0.977	.5989105	1.645121
	Black	1.09941	.0675755	1.54	0.123	.9746312	1.240163
	Latino	1.436886	.0809047	6.44	0.000	1.286752	1.604537
	Other	.8163482	.1645088	-1.01	0.314	.5499751	1.211736
	AsianMale	1.634154	.6945542	1.16	0.248	.7104151	3.759013
	BlackMale	1.727229	.1707779	5.53	0.000	1.422945	2.09658
	LatinoMale	1.165883	.0949038	1.89	0.059	.9939541	1.367551
	OtherMale	1.825738	.5571165	1.97	0.049	1.003917	3.320314
3							
	tenure	1.013884	.0023554	5.94	0.000	1.009278	1.018511
	exper	1.061667	.0036333	17.49	0.000	1.054569	1.068812
	male	1.394464	.0592052	7.83	0.000	1.283121	1.515469
	age	.9733836	.0029499	-8.90	0.000	.967619	.9791825
	pyacctrati~2	1.467503	.0465671	12.09	0.000	1.379013	1.56167
	pyperstudlep	1.011297	.0014005	8.11	0.000	1.008555	1.014045
	pyperstude~s	1.020373	.0013854	14.85	0.000	1.017661	1.023092
	pyperstudw~e	1.008854	.0040777	2.18	0.029	1.000894	1.016878
	pyperstudb~k	.9935413	.004523	-1.42	0.155	.9847159	1.002446
	pyperstudl~o	.9883987	.0042428	-2.72	0.007	.9801178	.9967496
	pyperstudm~e	.9932116	.0024965	-2.71	0.007	.9883305	.9981168
	Asian	.7436645	.2441642	-0.90	0.367	.3907561	1.4153
	Black	.7678708	.0560458	-3.62	0.000	.6655184	.8859642
	Latino	1.244533	.0779104	3.49	0.000	1.100828	1.406998
	Other	.6022585	.1469929	-2.08	0.038	.3732765	.9717067
	AsianMale	.724613	.4471825	-0.52	0.602	.2161731	2.428905



BlackMale		1.63735	.1908882	4.23	0.000	1.302882	2.057679
LatinoMale		.9657745	.0881495	-0.38	0.703	.8075765	1.154962
OtherMale		1.359033	.5198346	0.80	0.423	.6421593	2.876189

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=23586)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----+-----				
tenure		40.698	2	0.000
exper		341.525	2	0.000
male		64.930	2	0.000
age		153.406	2	0.000
pyacctrati~2		163.138	2	0.000
pyperstudlep		80.445	2	0.000
pyperstude~s		396.346	2	0.000
pyperstudw~e		9.781	2	0.008
pyperstudb~k		2.366	2	0.306
pyperstudl~o		7.881	2	0.019
pyperstudm~e		19.131	2	0.000
Asian		0.924	2	0.630
Black		21.530	2	0.000
Latino		42.900	2	0.000
Other		4.822	2	0.090
AsianMale		2.241	2	0.326
BlackMale		36.087	2	0.000
LatinoMale		5.136	2	0.077
OtherMale		3.962	2	0.138

\*\*\*\* Wald tests for independent variables (N=23586)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----+-----				
tenure		40.652	2	0.000
exper		319.670	2	0.000
male		64.819	2	0.000
age		151.287	2	0.000
pyacctrati~2		164.507	2	0.000
pyperstudlep		79.448	2	0.000
pyperstude~s		384.153	2	0.000
pyperstudw~e		9.616	2	0.008
pyperstudb~k		2.393	2	0.302
pyperstudl~o		8.071	2	0.018
pyperstudm~e		18.739	2	0.000
Asian		0.873	2	0.646
Black		21.099	2	0.000
Latino		42.849	2	0.000
Other		4.563	2	0.102
AsianMale		2.215	2	0.330
BlackMale		35.471	2	0.000
LatinoMale		5.140	2	0.077
OtherMale		3.927	2	0.140

\*\*\*\* Hausman tests of IIA assumption (N=23586)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
-----+-----					

```

      2 | -137.510  19    ---    ---
      3 |  -2.360   19    ---    ---
-----
Note: If chi2<0, the estimated model does not
meet asymptotic assumptions of the test.

**** Small-Hsiao tests of IIA assumption (N=23586)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted | lnL(full) lnL(omit) chi2 df P>chi2 evidence
-----+-----
      2 | -5112.949 -5102.385 21.129 20 0.390 for Ho
      3 | -5846.536 -5830.470 32.131 20 0.042 against Ho
-----

**** Wald tests for combining alternatives (N=23586)

Ho: All coefficients except intercepts associated with a given pair
of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested| chi2 df P>chi2
-----+-----
      2-      3 | 1491.442 19 0.000
      2-      1 | 511.435 19 0.000
      3-      1 | 1373.114 19 0.000
-----

**** LR tests for combining alternatives (N=23586)

Ho: All coefficients except intercepts associated with a given pair
of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested| chi2 df P>chi2
-----+-----
      2-      3 | 1660.404 19 0.000
      2-      1 | 528.174 19 0.000
      3-      1 | 1523.578 19 0.000
-----

by region:

. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
> AsianMale BlackMale LatinoMale OtherMale if validcert2006a==1, rrr iter(20)

-----> region = 1

Iteration 0: log likelihood = -1764.9169
Iteration 1: log likelihood = -1671.7365
Iteration 2: log likelihood = -1668.4124
Iteration 3: log likelihood = -1668.1019
Iteration 4: log likelihood = -1668.0524
Iteration 5: log likelihood = -1668.0473
Iteration 6: log likelihood = -1668.0462
Iteration 7: log likelihood = -1668.0459
Iteration 8: log likelihood = -1668.0459

Multinomial logistic regression          Number of obs   =      1636
                                         LR chi2(38)      =      193.74
                                         Prob > chi2      =      0.0000
Log likelihood = -1668.0459             Pseudo R2       =      0.0549
-----

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----							
1							
	tenure	.9858921	.0087293	-1.60	0.109	.9689306	1.003151
	exper	1.002602	.0131739	0.20	0.843	.977111	1.028758
	male	.7757454	.232492	-0.85	0.397	.4311346	1.395808
	age	1.026722	.0114223	2.37	0.018	1.004577	1.049355
pyacctrati~2		1.22731	.1612265	1.56	0.119	.9487143	1.587717
pyperstudlep		.9943531	.003124	-1.80	0.071	.988249	1.000495
pyperstude~s		.990617	.0062946	-1.48	0.138	.9783563	1.003031
pyperstudw~e		1.001752	.070964	0.02	0.980	.8718887	1.150957
pyperstudb~k		1.025606	.0858408	0.30	0.763	.8704363	1.208437
pyperstudl~o		1.00866	.0642803	0.14	0.892	.8902235	1.142853
pyperstudm~e		1.010563	.0091998	1.15	0.248	.9926919	1.028756
	Asian	3154634	3.98e+09	0.01	0.991	0	.
	Black	816520.9	8.32e+08	0.01	0.989	0	.
	Latino	.9980666	.2164476	-0.01	0.993	.6524723	1.526711
	Other	.6233236	.4934443	-0.60	0.550	.132091	2.941398
	AsianMale	7.36e-14	1.30e-10	-0.02	0.986	0	.
	BlackMale	3.65e-06	.0037252	-0.01	0.990	0	.
	LatinoMale	.8111706	.2660833	-0.64	0.523	.4264792	1.54286
	OtherMale	2.144287	8644.046	0.00	1.000	0	.
-----							
2		(base outcome)					
-----							
3							
	tenure	1.010419	.0100008	1.05	0.295	.9910068	1.030212
	exper	1.108954	.0205185	5.59	0.000	1.069459	1.149907
	male	.650764	.2405404	-1.16	0.245	.3153522	1.342923
	age	.9463435	.015616	-3.34	0.001	.9162263	.9774507
pyacctrati~2		1.43527	.2069065	2.51	0.012	1.081996	1.903889
pyperstudlep		1.011524	.0035227	3.29	0.001	1.004643	1.018452
pyperstude~s		.9938826	.0077985	-0.78	0.434	.9787147	1.009286
pyperstudw~e		1.126419	.1117614	1.20	0.230	.9273534	1.368216
pyperstudb~k		1.127813	.1459851	0.93	0.353	.8750981	1.453508
pyperstudl~o		1.124969	.0993866	1.33	0.183	.9461074	1.337645
pyperstudm~e		.9933083	.0120047	-0.56	0.579	.970056	1.017118
	Asian	1.420843	2830.855	0.00	1.000	0	.
	Black	2812961	2.87e+09	0.01	0.988	0	.
	Latino	.7793739	.1914072	-1.01	0.310	.4816132	1.261227
	Other	.3122481	.3634635	-1.00	0.317	.0318921	3.057149
	AsianMale	5.91e-07	.0014924	-0.01	0.995	0	.
	BlackMale	1.33e-12	1.67e-09	-0.02	0.983	0	.
	LatinoMale	1.302238	.5210997	0.66	0.509	.5943966	2.853019
	OtherMale	1.42e+08	4.04e+11	0.01	0.995	0	.
-----							

-----> region = 2

```
Iteration 0: log likelihood = -757.15308
Iteration 1: log likelihood = -711.24781
Iteration 2: log likelihood = -709.23815
Iteration 3: log likelihood = -709.01948
Iteration 4: log likelihood = -708.98197
Iteration 5: log likelihood = -708.97361
Iteration 6: log likelihood = -708.97158
Iteration 7: log likelihood = -708.97115
Iteration 8: log likelihood = -708.97107
Iteration 9: log likelihood = -708.97104
```

```
Multinomial logistic regression      Number of obs   =      740
                                      LR chi2(38)       =      96.36
                                      Prob > chi2       =      0.0000
Log likelihood = -708.97104          Pseudo R2       =      0.0636
```

-----

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.023812	.0162371	1.48	0.138	.9924774	1.056136
exper	.9732148	.0186146	-1.42	0.156	.9374063	1.010391
male	1.854362	.5346771	2.14	0.032	1.053813	3.263064
age	.9776105	.0149243	-1.48	0.138	.9487926	1.007304
pyacctrati~2	.8405803	.1788219	-0.82	0.414	.5539847	1.275442
pyperstudlep	.9716545	.0154774	-1.81	0.071	.941788	1.002468
pyperstude~s	1.009692	.0080601	1.21	0.227	.9940171	1.025613
pyperstudw~e	.9281179	.0710056	-0.98	0.330	.7988809	1.078262
pyperstudb~k	.9041989	.0797339	-1.14	0.253	.7606826	1.074792
pyperstudl~o	.9268006	.0690318	-1.02	0.307	.8009132	1.072475
pyperstudm~e	1.013854	.0116372	1.20	0.231	.9913004	1.036921
Asian	1.04e-06	.0021051	-0.01	0.995	0	.
Black	3.635262	3.150671	1.49	0.136	.6649696	19.87328
Latino	1.681316	.4455926	1.96	0.050	1.000135	2.826441
Other	1.34e-06	.0011684	-0.02	0.988	0	.
AsianMale	1.011122	2472.652	0.00	1.000	0	.
BlackMale	.8217178	1.057903	-0.15	0.879	.0658963	10.2467
LatinoMale	.4967531	.1998367	-1.74	0.082	.2257951	1.092866
OtherMale	1432487	1.25e+09	0.02	0.987	0	.
3						
tenure	.9943856	.0125058	-0.45	0.654	.9701743	1.019201
exper	1.035864	.0177506	2.06	0.040	1.001652	1.071246
male	.9346844	.2704769	-0.23	0.815	.5300859	1.648101
age	1.001058	.0153903	0.07	0.945	.9713439	1.031682
pyacctrati~2	1.583503	.3086162	2.36	0.018	1.080748	2.320137
pyperstudlep	1.010195	.0136718	0.75	0.454	.9837515	1.03735
pyperstude~s	1.026633	.0082874	3.26	0.001	1.010518	1.043005
pyperstudw~e	1.078368	.084509	0.96	0.336	.9248265	1.2574
pyperstudb~k	1.061873	.094922	0.67	0.502	.8912153	1.265209
pyperstudl~o	1.057187	.0807902	0.73	0.467	.9101294	1.228007
pyperstudm~e	.9615469	.0144176	-2.62	0.009	.9337	.9902243
Asian	1.62e-06	.0031249	-0.01	0.994	0	.
Black	4.220299	3.377135	1.80	0.072	.8794299	20.25281
Latino	.874483	.2234883	-0.52	0.600	.5299233	1.443078
Other	.9002692	1.071871	-0.09	0.930	.0872815	9.285873
AsianMale	.4082813	962.3273	-0.00	1.000	0	.
BlackMale	.4672612	.7017137	-0.51	0.612	.0246184	8.868689
LatinoMale	1.998867	.7925448	1.75	0.081	.9189356	4.347932
OtherMale	1.71e-06	.0021979	-0.01	0.992	0	.

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -317.95378  
Iteration 1: log likelihood = -285.41561  
Iteration 2: log likelihood = -284.39815  
Iteration 3: log likelihood = -284.34119  
Iteration 4: log likelihood = -284.33247  
Iteration 5: log likelihood = -284.33036  
Iteration 6: log likelihood = -284.32992  
Iteration 7: log likelihood = -284.32982  
Iteration 8: log likelihood = -284.3298

Multinomial logistic regression	Number of obs	=	295
	LR chi2 (32)	=	67.25
	Prob > chi2	=	0.0003
Log likelihood = -284.3298	Pseudo R2	=	0.1058

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.056928	.0265554	2.20	0.028	1.006142	1.110279
	exper	.9309113	.0263131	-2.53	0.011	.8807412	.9839392
	male	2.162798	.8478747	1.97	0.049	1.003049	4.663476
	age	.9752893	.0239909	-1.02	0.309	.9293835	1.023463
pyacctrati~2		1.345934	.54132	0.74	0.460	.6118989	2.960518
pyperstudlep		1.024309	.0352019	0.70	0.485	.9575872	1.095681
pyperstude~s		.9867275	.0180506	-0.73	0.465	.9519758	1.022748
pyperstudw~e		.9737622	.0831459	-0.31	0.756	.8237056	1.151155
pyperstudb~k		.9975445	.0868004	-0.03	0.977	.8411351	1.183038
pyperstudl~o		.9968214	.0883738	-0.04	0.971	.8378255	1.18599
pyperstudm~e		1.028596	.026873	1.08	0.281	.9772516	1.082638
	Asian	(omitted)					
	Black	2.932709	2.488871	1.27	0.205	.5557622	15.47565
	Latino	1.835111	1.077161	1.03	0.301	.580804	5.798223
	Other	616152.4	2.52e+08	0.03	0.974	0	.
	AsianMale	(omitted)					
	BlackMale	1.396752	2.081846	0.22	0.823	.0752335	25.93149
	LatinoMale	.5055377	.4568892	-0.75	0.450	.0859932	2.97196
	OtherMale	(omitted)					
3							
	tenure	1.020737	.0198787	1.05	0.292	.98251	1.060452
	exper	1.040744	.0300495	1.38	0.167	.9834837	1.101339
	male	.997783	.3465082	-0.01	0.995	.5051641	1.970787
	age	.9386475	.0263623	-2.25	0.024	.8883748	.9917652
pyacctrati~2		1.603425	.517964	1.46	0.144	.8512947	3.020073
pyperstudlep		1.002242	.0327316	0.07	0.945	.9400991	1.068492
pyperstude~s		1.037064	.0169334	2.23	0.026	1.0044	1.07079
pyperstudw~e		.9748495	.0700748	-0.35	0.723	.8467417	1.122339
pyperstudb~k		.9433177	.069304	-0.79	0.427	.8168111	1.089417
pyperstudl~o		.9410785	.0708398	-0.81	0.420	.8119916	1.090687
pyperstudm~e		1.000583	.0274128	0.02	0.983	.9482718	1.05578
	Asian	(omitted)					
	Black	2.397136	1.832702	1.14	0.253	.5356939	10.72676
	Latino	.9000852	.5617151	-0.17	0.866	.2648959	3.058384
	Other	1136289	4.64e+08	0.03	0.973	0	.
	AsianMale	(omitted)					
	BlackMale	1.838506	2.701209	0.41	0.679	.1032391	32.74053
	LatinoMale	2.609888	2.298495	1.09	0.276	.4644934	14.66439
	OtherMale	(omitted)					

-----> region = 4

```
Iteration 0: log likelihood = -5278.0189
Iteration 1: log likelihood = -4982.3946
Iteration 2: log likelihood = -4968.9374
Iteration 3: log likelihood = -4968.6946
Iteration 4: log likelihood = -4968.6384
Iteration 5: log likelihood = -4968.627
Iteration 6: log likelihood = -4968.6245
Iteration 7: log likelihood = -4968.624
Iteration 8: log likelihood = -4968.6239
Iteration 9: log likelihood = -4968.6238
```

Multinomial logistic regression

```
Number of obs   =      5145
LR chi2(38)     =      618.79
Prob > chi2     =      0.0000
Pseudo R2      =      0.0586
```

Log likelihood = -4968.6238

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.014506	.005382	2.71	0.007	1.004012	1.02511
	exper	1.033843	.0071608	4.81	0.000	1.019903	1.047974
	male	1.098118	.0990299	1.04	0.299	.9202087	1.310423
	age	.9613185	.0054352	-6.98	0.000	.9507244	.9720306
pyacctrati~2		1.080435	.0664564	1.26	0.208	.957728	1.218864
pyperstudlep		1.003574	.0029593	1.21	0.226	.9977905	1.009391
pyperstude~s		.9921012	.003447	-2.28	0.022	.9853682	.9988803
pyperstudw~e		1.006653	.0053494	1.25	0.212	.9962224	1.017192
pyperstudb~k		1.008519	.0066527	1.29	0.198	.9955639	1.021643
pyperstudl~o		1.008079	.0064184	1.26	0.206	.9955772	1.020738
pyperstudm~e		1.008499	.0056253	1.52	0.129	.9975339	1.019585
	Asian	.9845184	.3751562	-0.04	0.967	.4665184	2.077681
	Black	.9106645	.0945742	-0.90	0.368	.7429499	1.116239
	Latino	1.451571	.1902825	2.84	0.004	1.122681	1.876808
	Other	.9053109	.3409525	-0.26	0.792	.4327366	1.893965
	AsianMale	1.245036	.9414269	0.29	0.772	.2828464	5.480411
	BlackMale	2.076159	.3450491	4.40	0.000	1.498975	2.875589
	LatinoMale	1.17748	.2508764	0.77	0.443	.775523	1.787773
	OtherMale	1.664586	1.204588	0.70	0.481	.403022	6.87517
3							
	tenure	1.035655	.00612	5.93	0.000	1.023729	1.047719
	exper	1.074837	.0093724	8.28	0.000	1.056624	1.093365
	male	1.131526	.1242782	1.13	0.261	.9123789	1.403312
	age	.9692695	.0072236	-4.19	0.000	.9552144	.9835314
pyacctrati~2		1.349118	.0985684	4.10	0.000	1.169122	1.556825
pyperstudlep		1.020467	.0036473	5.67	0.000	1.013343	1.027641
pyperstude~s		1.020777	.0045177	4.65	0.000	1.011961	1.02967
pyperstudw~e		1.012948	.0067634	1.93	0.054	.9997784	1.026291
pyperstudb~k		.9987922	.0084689	-0.14	0.887	.9823306	1.01553
pyperstudl~o		.9821836	.0079091	-2.23	0.026	.9668038	.9978081
pyperstudm~e		1.000721	.0071979	0.10	0.920	.9867121	1.014928
	Asian	.6563255	.3464199	-0.80	0.425	.2332601	1.846707
	Black	.4537383	.0633872	-5.66	0.000	.3450584	.5966481
	Latino	1.285303	.2024928	1.59	0.111	.94385	1.750281
	Other	.6769319	.3497434	-0.76	0.450	.2459028	1.863487
	AsianMale	2.57e-06	.0011832	-0.03	0.978	0	.
	BlackMale	2.31862	.5067696	3.85	0.000	1.510729	3.558546
	LatinoMale	1.177836	.3165913	0.61	0.543	.6954881	1.99471
	OtherMale	3.284904	2.939986	1.33	0.184	.5684576	18.98223

-----> region = 5

note: OtherMale omitted because of collinearity

Iteration 0: log likelihood = -550.04507  
Iteration 1: log likelihood = -504.91476  
Iteration 2: log likelihood = -503.06156  
Iteration 3: log likelihood = -502.88644  
Iteration 4: log likelihood = -502.85725  
Iteration 5: log likelihood = -502.85067  
Iteration 6: log likelihood = -502.84906  
Iteration 7: log likelihood = -502.84874  
Iteration 8: log likelihood = -502.84867  
Iteration 9: log likelihood = -502.84865

Multinomial logistic regression

Number of obs = 545  
LR chi2(36) = 94.39  
Prob > chi2 = 0.0000

Log likelihood = -502.84865

Pseudo R2 = 0.0858

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.015704	.0167544	0.94	0.345	.9833915	1.049079
exper	.997286	.0230214	-0.12	0.906	.9531704	1.043443
male	2.721976	.77028	3.54	0.000	1.563171	4.739822
age	.977071	.0190439	-1.19	0.234	.9404495	1.015118
pyacctrati~2	.877097	.2214039	-0.52	0.603	.5347865	1.438516
pyperstudlep	.9418584	.0350608	-1.61	0.108	.8755875	1.013145
pyperstude~s	1.014236	.0101826	1.41	0.159	.9944737	1.034391
pyperstudw~e	.9524691	.0451664	-1.03	0.304	.8679339	1.045238
pyperstudb~k	.9452494	.0466363	-1.14	0.254	.8581244	1.04122
pyperstudl~o	.9588572	.0559736	-0.72	0.472	.8551943	1.075086
pyperstudm~e	.9826433	.0224047	-0.77	0.443	.9396977	1.027552
Asian	7.61e-07	.0008735	-0.01	0.990	0	.
Black	1.760257	.695307	1.43	0.152	.811613	3.817714
Latino	.957505	1.129783	-0.04	0.971	.0947955	9.671509
Other	9.27e-07	.0018418	-0.01	0.994	0	.
AsianMale	1230498	1.41e+09	0.01	0.990	0	.
BlackMale	1.251659	.6762638	0.42	0.678	.4340971	3.608985
LatinoMale	3.401123	5.826169	0.71	0.475	.1184412	97.66565
OtherMale	(omitted)					
3						
tenure	1.014277	.0147611	0.97	0.330	.985755	1.043625
exper	1.058481	.0258562	2.33	0.020	1.008998	1.110391
male	2.17547	.6121416	2.76	0.006	1.253253	3.776306
age	.9773051	.0210424	-1.07	0.286	.9369209	1.01943
pyacctrati~2	1.840545	.4312322	2.60	0.009	1.162823	2.913262
pyperstudlep	.9887629	.0311636	-0.36	0.720	.9295317	1.051768
pyperstude~s	1.022003	.010198	2.18	0.029	1.002209	1.042188
pyperstudw~e	1.016272	.0437444	0.37	0.708	.9340517	1.10573
pyperstudb~k	.9955734	.0447404	-0.10	0.921	.9116347	1.087241
pyperstudl~o	1.004642	.0579065	0.08	0.936	.8973237	1.124796
pyperstudm~e	.9829428	.026697	-0.63	0.526	.931986	1.036686
Asian	2.35e-06	.0025234	-0.01	0.990	0	.
Black	2.097845	.8938672	1.74	0.082	.910094	4.835712
Latino	1.161452	1.432495	0.12	0.903	.103552	13.02699
Other	2.84e-06	.0056218	-0.01	0.995	0	.
AsianMale	.5990122	986.2855	-0.00	1.000	0	.
BlackMale	2.007736	1.159898	1.21	0.228	.6470759	6.229566
LatinoMale	2.126373	4.076974	0.39	0.694	.0496134	91.13391
OtherMale	(omitted)					

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -867.18625  
 Iteration 1: log likelihood = -807.36623  
 Iteration 2: log likelihood = -805.10611  
 Iteration 3: log likelihood = -804.97476  
 Iteration 4: log likelihood = -804.94605  
 Iteration 5: log likelihood = -804.94002  
 Iteration 6: log likelihood = -804.93905  
 Iteration 7: log likelihood = -804.93881  
 Iteration 8: log likelihood = -804.93876  
 Iteration 9: log likelihood = -804.93875

Multinomial logistic regression

Number of obs = 810  
 LR chi2(36) = 124.50

Log likelihood = -804.93875      Prob > chi2 = 0.0000  
Pseudo R2 = 0.0718

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9921973	.0137937	-0.56	0.573	.9655271	1.019604
	exper	1.027842	.0164911	1.71	0.087	.9960235	1.060678
	male	1.54659	.2982561	2.26	0.024	1.059798	2.256978
	age	.959602	.0129545	-3.05	0.002	.9345445	.9853313
pyacctrati~2		1.230747	.1950384	1.31	0.190	.9021471	1.679036
pyperstudlep		1.00834	.0170014	0.49	0.622	.9755623	1.042219
pyperstude~s		1.002158	.0072423	0.30	0.766	.9880629	1.016453
pyperstudw~e		1.013078	.0335158	0.39	0.695	.9494727	1.080944
pyperstudb~k		.9983585	.0349107	-0.05	0.963	.9322268	1.069182
pyperstudl~o		1.007516	.0337997	0.22	0.823	.9434007	1.075988
pyperstudm~e		.9809187	.0173568	-1.09	0.276	.9474832	1.015534
	Asian	2.853467	4.121834	0.73	0.468	.1681946	48.40985
	Black	1.475392	.6839446	0.84	0.401	.5947271	3.660136
	Latino	.9158245	.5385192	-0.15	0.881	.2892627	2.89956
	Other	9.78e-07	.0013267	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	1.789996	1.401307	0.74	0.457	.3859088	8.302702
	LatinoMale	1.024106	.8227876	0.03	0.976	.2120665	4.945584
	OtherMale	1408147	1.91e+09	0.01	0.992	0	.
3							
	tenure	.9938958	.013262	-0.46	0.646	.9682397	1.020232
	exper	1.11225	.0221559	5.34	0.000	1.069662	1.156534
	male	2.256615	.4648752	3.95	0.000	1.506972	3.379167
	age	.9251617	.0164894	-4.36	0.000	.8934011	.9580514
pyacctrati~2		1.642682	.2774768	2.94	0.003	1.179699	2.287367
pyperstudlep		1.044209	.0183603	2.46	0.014	1.008837	1.080822
pyperstude~s		1.028336	.0079879	3.60	0.000	1.012798	1.044112
pyperstudw~e		1.0194	.0355416	0.55	0.582	.9520662	1.091495
pyperstudb~k		1.010384	.0364638	0.29	0.775	.9413856	1.08444
pyperstudl~o		.9885049	.0348327	-0.33	0.743	.9225382	1.059188
pyperstudm~e		.9591923	.0202427	-1.97	0.048	.9203266	.9996993
	Asian	2.64e-06	.0029679	-0.01	0.991	0	.
	Black	1.871291	.8853781	1.32	0.185	.7402969	4.73017
	Latino	.8363183	.5631054	-0.27	0.791	.2234833	3.129667
	Other	1.27e-06	.001815	-0.01	0.992	0	.
	AsianMale	(omitted)					
	BlackMale	.7736286	.6566471	-0.30	0.762	.1465691	4.083406
	LatinoMale	.2071909	.2634493	-1.24	0.216	.0171413	2.50437
	OtherMale	.6281561	1128.364	-0.00	1.000	0	.

-----> region = 7

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1049.3548  
Iteration 1: log likelihood = -988.66374  
Iteration 2: log likelihood = -986.9767  
Iteration 3: log likelihood = -986.84966  
Iteration 4: log likelihood = -986.82295  
Iteration 5: log likelihood = -986.81708  
Iteration 6: log likelihood = -986.81612  
Iteration 7: log likelihood = -986.81601  
Iteration 8: log likelihood = -986.81599

Multinomial logistic regression      Number of obs = 1019



Log likelihood = -986.81599

LR chi2(34) = 125.08  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0596

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	.9807014	.0125703	-1.52	0.128	.956371 1.005651
	exper	1.000672	.0160512	0.04	0.967	.9697013 1.032631
	male	.997944	.1969489	-0.01	0.992	.677825 1.469247
	age	.9777452	.0137822	-1.60	0.110	.9511023 1.005134
	pyacctrati~2	1.053827	.2195712	0.25	0.801	.7005134 1.58534
	pyperstudlep	.9795446	.0167605	-1.21	0.227	.9472393 1.012952
	pyperstude~s	.985449	.0081817	-1.77	0.077	.969543 1.001616
	pyperstudw~e	.9010708	.0668205	-1.40	0.160	.7791777 1.042032
	pyperstudb~k	.9147414	.0692732	-1.18	0.239	.788564 1.061108
	pyperstudl~o	.9227361	.069496	-1.07	0.286	.7961027 1.069513
	pyperstudm~e	.9963929	.0127703	-0.28	0.778	.9716752 1.021739
	Asian (omitted)					
	Black	1.226205	.4364836	0.57	0.567	.6103334 2.463538
	Latino	.8490034	1.014351	-0.14	0.891	.0816456 8.828486
	Other	2.26e-06	.0015177	-0.02	0.985	0 .
	AsianMale (omitted)					
	BlackMale	1.954435	.9905112	1.32	0.186	.7238203 5.277299
	LatinoMale	4.020175	6.124526	0.91	0.361	.2029945 79.61699
	OtherMale	.6419089	793.762	-0.00	1.000	0 .
3						
	tenure	.9918432	.0099105	-0.82	0.412	.9726079 1.011459
	exper	1.036683	.0143541	2.60	0.009	1.008928 1.065202
	male	1.651298	.2800853	2.96	0.003	1.184264 2.302513
	age	.9885347	.0124717	-0.91	0.361	.9643905 1.013283
	pyacctrati~2	1.427796	.2289578	2.22	0.026	1.042724 1.955072
	pyperstudlep	1.01883	.015025	1.26	0.206	.9898032 1.048708
	pyperstude~s	1.031871	.0070929	4.56	0.000	1.018062 1.045867
	pyperstudw~e	1.010344	.0842226	0.12	0.902	.8580503 1.189668
	pyperstudb~k	1.002297	.0847007	0.03	0.978	.8493065 1.182847
	pyperstudl~o	.9768722	.0825856	-0.28	0.782	.8277067 1.15292
	pyperstudm~e	.9883898	.0108778	-1.06	0.289	.967298 1.009942
	Asian (omitted)					
	Black	1.209792	.3981872	0.58	0.563	.6346709 2.30607
	Latino	.7578727	.8882621	-0.24	0.813	.0761989 7.537792
	Other	1.231296	1.53778	0.17	0.868	.1064848 14.23763
	AsianMale (omitted)					
	BlackMale	1.254381	.5997157	0.47	0.635	.4914426 3.201742
	LatinoMale	6.481011	9.325173	1.30	0.194	.3862739 108.7402
	OtherMale	9.11e-07	.0007111	-0.02	0.986	0 .

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -395.49344  
 Iteration 1: log likelihood = -352.99466  
 Iteration 2: log likelihood = -351.80287  
 Iteration 3: log likelihood = -351.74569  
 Iteration 4: log likelihood = -351.73254  
 Iteration 5: log likelihood = -351.72963  
 Iteration 6: log likelihood = -351.72898  
 Iteration 7: log likelihood = -351.72888  
 Iteration 8: log likelihood = -351.72887

Multinomial logistic regression Number of obs = 384

Log likelihood = -351.72887

LR chi2(36)	=	87.53
Prob > chi2	=	0.0000
Pseudo R2	=	0.1107

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.012316	.0237477	0.52	0.602	.966825	1.059947
	exper	1.009131	.0317852	0.29	0.773	.9487169	1.073392
	male	2.328044	.7836889	2.51	0.012	1.203514	4.503301
	age	.9456281	.0275338	-1.92	0.055	.8931737	1.001163
pyacctrati~2		1.335731	.4617875	0.84	0.402	.6783326	2.630241
pyperstudlep		.9369333	.0465543	-1.31	0.190	.8499908	1.032769
pyperstude~s		.9964937	.0138308	-0.25	0.800	.9697512	1.023974
pyperstudw~e		.7485874	.1235758	-1.75	0.079	.5416607	1.034565
pyperstudb~k		.768707	.1275889	-1.58	0.113	.5552385	1.064246
pyperstudl~o		.7825034	.1305409	-1.47	0.142	.5642672	1.085145
pyperstudm~e		1.028555	.0273711	1.06	0.290	.9762833	1.083625
	Asian	2.49865	5812.765	0.00	1.000	0	.
	Black	2.175178	1.2446	1.36	0.174	.7086825	6.676332
	Latino	9.735677	14.86243	1.49	0.136	.4885699	194.0017
	Other	4.753586	7.085471	1.05	0.296	.256012	88.26375
	AsianMale	(omitted)					
	BlackMale	.8770759	.9942102	-0.12	0.908	.0950948	8.089422
	LatinoMale	.1276672	297.0003	-0.00	0.999	0	.
	OtherMale	1781763	4.64e+09	0.01	0.996	0	.
3							
	tenure	1.007666	.01822	0.42	0.673	.972581	1.044017
	exper	1.013072	.0248675	0.53	0.597	.9654861	1.063002
	male	2.540957	.7086688	3.34	0.001	1.470952	4.389308
	age	1.010341	.0236099	0.44	0.660	.9651097	1.057691
pyacctrati~2		2.001145	.5551383	2.50	0.012	1.16184	3.446757
pyperstudlep		1.012942	.033075	0.39	0.694	.9501472	1.079887
pyperstude~s		1.045137	.0122146	3.78	0.000	1.021469	1.069353
pyperstudw~e		.918615	.1293897	-0.60	0.547	.6970103	1.210676
pyperstudb~k		.893035	.1259131	-0.80	0.422	.6774133	1.17729
pyperstudl~o		.8858369	.1251301	-0.86	0.391	.6716084	1.1684
pyperstudm~e		1.033033	.0260089	1.29	0.197	.9832939	1.085288
	Asian	563616.4	6.93e+08	0.01	0.991	0	.
	Black	1.599084	.9024613	0.83	0.406	.529037	4.833443
	Latino	4.881176	6.216002	1.24	0.213	.4022998	59.22419
	Other	3.314434	4.771037	0.83	0.405	.1973006	55.67888
	AsianMale	(omitted)					
	BlackMale	1.150944	1.228314	0.13	0.895	.1421124	9.321298
	LatinoMale	408196.1	5.02e+08	0.01	0.992	0	.
	OtherMale	.136508	564.6848	-0.00	1.000	0	.

-----> region = 9

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity

Iteration 0:	log likelihood = -225.13938
Iteration 1:	log likelihood = -201.5717
Iteration 2:	log likelihood = -200.67438
Iteration 3:	log likelihood = -199.57327
Iteration 4:	log likelihood = -199.50216
Iteration 5:	log likelihood = -199.48874
Iteration 6:	log likelihood = -199.48587
Iteration 7:	log likelihood = -199.48517
Iteration 8:	log likelihood = -199.48502
Iteration 9:	log likelihood = -199.48499

Iteration 10: log likelihood = -199.48498

Multinomial logistic regression

Number of obs = 213  
 LR chi2(34) = 51.31  
 Prob > chi2 = 0.0288  
 Pseudo R2 = 0.1139

Log likelihood = -199.48498

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	tenure	.9754088	.0230774	-1.05	0.293	.9312106	1.021705
	exper	.9776568	.0335122	-0.66	0.510	.914132	1.045596
	male	.7430556	.2677474	-0.82	0.410	.3666953	1.505696
	age	1.018765	.0293662	0.64	0.519	.9628037	1.077978
	pyacctrati~2	.6949108	.2354976	-1.07	0.283	.3576566	1.350181
	pyperstudlep	1.010102	.0621301	0.16	0.870	.8953831	1.139519
	pyperstude~s	.9885005	.015295	-0.75	0.455	.9589729	1.018937
	pyperstudw~e	.8047641	.0906686	-1.93	0.054	.6453095	1.00362
	pyperstudb~k	.8162259	.0990438	-1.67	0.094	.6434613	1.035376
	pyperstudl~o	.803749	.091278	-1.92	0.054	.6433592	1.004124
	pyperstudm~e	.949732	.0318031	-1.54	0.124	.8894007	1.014156
	Asian	.8441708	7829.185	-0.00	1.000	0	.
	Black	5.34e-07	.0008773	-0.01	0.993	0	.
	Latino	2.87e-07	.0005017	-0.01	0.993	0	.
	Other	5488774	1.48e+10	0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	1.57e+13	4.63e+16	0.01	0.992	0	.
	LatinoMale	(omitted)					
	OtherMale	3.94e-14	1.16e-10	-0.01	0.992	0	.
2							
	tenure	.9736838	.0300384	-0.86	0.387	.9165542	1.034374
	exper	1.006333	.0447023	0.14	0.887	.9224242	1.097875
	male	.7198203	.3221954	-0.73	0.463	.2993793	1.730719
	age	.9731462	.0351716	-0.75	0.451	.906596	1.044582
	pyacctrati~2	.7910482	.3390128	-0.55	0.584	.3415158	1.832294
	pyperstudlep	1.026319	.0853401	0.31	0.755	.8719739	1.207984
	pyperstude~s	.9634793	.0200967	-1.78	0.074	.9248849	1.003684
	pyperstudw~e	.6980313	.0903029	-2.78	0.005	.5416974	.8994832
	pyperstudb~k	.732788	.1014122	-2.25	0.025	.5587	.961121
	pyperstudl~o	.7122188	.0931984	-2.59	0.010	.5510973	.9204466
	pyperstudm~e	.9943063	.0260803	-0.22	0.828	.9444816	1.046759
	Asian	1.14e+08	7.19e+11	0.00	0.998	0	.
	Black	1.325321	2.254314	0.17	0.868	.0472567	37.16879
	Latino	3.83e-07	.0009253	-0.01	0.995	0	.
	Other	1.427772	6736.381	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	3.11e+07	7.64e+10	0.01	0.994	0	.
	LatinoMale	(omitted)					
	OtherMale	.4068755	1919.682	-0.00	1.000	0	.
3		(base outcome)					

-----> region = 10

Iteration 0: log likelihood = -3730.9129  
 Iteration 1: log likelihood = -3496.027  
 Iteration 2: log likelihood = -3490.285  
 Iteration 3: log likelihood = -3490.2574  
 Iteration 4: log likelihood = -3490.2574

Multinomial logistic regression

Number of obs = 3487  
 LR chi2(38) = 481.31  
 Prob > chi2 = 0.0000

Log likelihood = -3490.2574

Pseudo R2 = 0.0645

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.018496	.0065414	2.85	0.004	1.005755	1.031398
exper	1.015213	.0084043	1.82	0.068	.9988738	1.03182
male	1.31738	.1397471	2.60	0.009	1.070079	1.621833
age	.9653802	.0065762	-5.17	0.000	.9525766	.9783558
pyacctrati~2	.9314429	.0719418	-0.92	0.358	.8005935	1.083678
pyperstudlep	1.005031	.0047597	1.06	0.289	.9957454	1.014403
pyperstude~s	.9916135	.0035031	-2.38	0.017	.9847713	.9985032
pyperstudw~e	.9821913	.0079105	-2.23	0.026	.9668088	.9978186
pyperstudb~k	.9846574	.0085825	-1.77	0.076	.9679789	1.001623
pyperstudl~o	.9843367	.0082846	-1.88	0.061	.9682325	1.000709
pyperstudm~e	1.007012	.005914	1.19	0.234	.995487	1.01867
Asian	3.589716	2.515345	1.82	0.068	.9091086	14.17439
Black	1.096027	.1467468	0.68	0.493	.8430512	1.424913
Latino	2.180856	.4075376	4.17	0.000	1.512037	3.145513
Other	.6758058	.2869	-0.92	0.356	.294077	1.55304
AsianMale	.3713593	.3549509	-1.04	0.300	.0570431	2.417607
BlackMale	1.572451	.337789	2.11	0.035	1.032107	2.395684
LatinoMale	1.141022	.3302721	0.46	0.649	.64701	2.012226
OtherMale	2.354131	1.398982	1.44	0.150	.7344963	7.545214
3						
tenure	1.018062	.0063956	2.85	0.004	1.005604	1.030675
exper	1.077806	.0103452	7.81	0.000	1.057719	1.098274
male	1.186719	.1390299	1.46	0.144	.9432469	1.493036
age	.9664705	.0081706	-4.03	0.000	.9505885	.9826179
pyacctrati~2	1.613132	.129287	5.97	0.000	1.378634	1.887517
pyperstudlep	1.020248	.0056211	3.64	0.000	1.00929	1.031325
pyperstude~s	1.014871	.0039511	3.79	0.000	1.007157	1.022645
pyperstudw~e	.9947287	.0086707	-0.61	0.544	.9778788	1.011869
pyperstudb~k	.9829172	.009527	-1.78	0.075	.9644209	1.001768
pyperstudl~o	.9667472	.0088518	-3.69	0.000	.9495528	.984253
pyperstudm~e	1.004913	.00708	0.70	0.487	.9911318	1.018886
Asian	3.022022	2.389137	1.40	0.162	.6417394	14.23104
Black	.8548918	.1381565	-0.97	0.332	.6228035	1.173468
Latino	2.774754	.5647973	5.01	0.000	1.861931	4.135094
Other	.3030915	.1748039	-2.07	0.038	.0978712	.9386259
AsianMale	.1887571	.2618094	-1.20	0.229	.0124529	2.861111
BlackMale	2.375808	.5960594	3.45	0.001	1.452968	3.884783
LatinoMale	1.153419	.3751111	0.44	0.661	.6097644	2.181785
OtherMale	1.649068	1.469959	0.56	0.575	.2874006	9.462137

-----&gt; region = 11

Iteration 0: log likelihood = -2643.7246  
Iteration 1: log likelihood = -2489.1754  
Iteration 2: log likelihood = -2485.0419  
Iteration 3: log likelihood = -2485.0259  
Iteration 4: log likelihood = -2485.0259

Multinomial logistic regression

Number of obs = 2454  
LR chi2(38) = 317.40  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0600

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
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1		(base outcome)					
2	tenure	1.001845	.007955	0.23	0.816	.9863738	1.017558
	exper	1.001647	.0094207	0.18	0.861	.9833524	1.020283
	male	1.32343	.1500457	2.47	0.013	1.059729	1.65275
	age	.9824116	.0074032	-2.35	0.019	.9680082	.9970293
	pyacctrati~2	1.15003	.1052017	1.53	0.126	.9612657	1.375862
	pyperstudlep	1.008896	.0065471	1.36	0.172	.9961451	1.02181
	pyperstude~s	.998658	.0040938	-0.33	0.743	.9906665	1.006714
	pyperstudw~e	.9979416	.0122394	-0.17	0.867	.9742389	1.022221
	pyperstudb~k	1.000176	.0138002	0.01	0.990	.9734906	1.027593
	pyperstudl~o	1.000313	.0132491	0.02	0.981	.974679	1.02662
	pyperstudm~e	.9977313	.0059103	-0.38	0.701	.9862144	1.009383
	Asian	.3505556	.3954745	-0.93	0.353	.0384137	3.199102
	Black	1.722144	.3540803	2.64	0.008	1.150956	2.576797
	Latino	1.74261	.4593758	2.11	0.035	1.03947	2.921384
	Other	.5115579	.2994538	-1.15	0.252	.1624133	1.611268
	AsianMale	5.974264	9.99438	1.07	0.285	.2250633	158.5858
	BlackMale	1.21657	.4236836	0.56	0.574	.6147475	2.40756
	LatinoMale	.762331	.2835036	-0.73	0.466	.3677815	1.580146
	OtherMale	.5874246	.581979	-0.54	0.591	.0842639	4.095084
3	tenure	1.000345	.0074713	0.05	0.963	.9858084	1.015096
	exper	1.08562	.0116599	7.65	0.000	1.063006	1.108715
	male	1.470099	.1828511	3.10	0.002	1.152057	1.875941
	age	.976603	.0091121	-2.54	0.011	.9589058	.9946267
	pyacctrati~2	1.440664	.1401485	3.75	0.000	1.190577	1.743284
	pyperstudlep	1.021302	.007658	2.81	0.005	1.006403	1.036422
	pyperstude~s	1.027237	.0048356	5.71	0.000	1.017803	1.036758
	pyperstudw~e	1.006689	.0136911	0.49	0.624	.9802092	1.033884
	pyperstudb~k	.991969	.0152985	-0.52	0.601	.9624331	1.022411
	pyperstudl~o	.9758166	.0145681	-1.64	0.101	.9476774	1.004791
	pyperstudm~e	.9786348	.0083367	-2.54	0.011	.9624307	.9951117
	Asian	.8841911	.8130077	-0.13	0.894	.1458385	5.360683
	Black	1.545091	.388311	1.73	0.083	.9441274	2.528584
	Latino	2.586873	.7398352	3.32	0.001	1.476845	4.531221
	Other	.4128047	.2804482	-1.30	0.193	.1090088	1.563247
	AsianMale	4.593082	7.139582	0.98	0.327	.2182623	96.65619
	BlackMale	1.174027	.4960754	0.38	0.704	.5128731	2.687486
	LatinoMale	.5913356	.2463878	-1.26	0.207	.2613182	1.33813
	OtherMale	1.735675	1.644203	0.58	0.561	.2710982	11.11247

-----> region = 12

Iteration 0: log likelihood = -878.18986  
Iteration 1: log likelihood = -816.8932  
Iteration 2: log likelihood = -815.16453  
Iteration 3: log likelihood = -814.88078  
Iteration 4: log likelihood = -814.81824  
Iteration 5: log likelihood = -814.80494  
Iteration 6: log likelihood = -814.80225  
Iteration 7: log likelihood = -814.80163  
Iteration 8: log likelihood = -814.80148  
Iteration 9: log likelihood = -814.80145

Multinomial logistic regression                      Number of obs    =        805  
   LR chi2(38)        =        126.78  
   Prob > chi2        =        0.0000  
Log likelihood = -814.80145                              Pseudo R2        =        0.0722

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
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1		(base outcome)					
2							
	tenure	1.003255	.0125626	0.26	0.795	.9789321	1.028181
	exper	1.017662	.0177101	1.01	0.314	.9835359	1.052972
	male	.6772162	.1430967	-1.84	0.065	.4475765	1.024678
	age	.9627244	.014948	-2.45	0.014	.9338682	.9924722
	pyacctrati~2	1.355288	.2505403	1.64	0.100	.94336	1.947089
	pyperstudlep	1.00432	.0184213	0.24	0.814	.9688565	1.041082
	pyperstude~s	.9912525	.0079538	-1.09	0.274	.9757853	1.006965
	pyperstudw~e	.956698	.046129	-0.92	0.359	.8704274	1.051519
	pyperstudb~k	.9820789	.0510902	-0.35	0.728	.8868799	1.087497
	pyperstudl~o	.9713091	.0470974	-0.60	0.548	.8832506	1.068147
	pyperstudm~e	1.013405	.0092485	1.46	0.145	.9954396	1.031695
	Asian	1.59e-06	.0015835	-0.01	0.989	0	.
	Black	1.047493	.3602287	0.13	0.893	.5338585	2.055303
	Latino	.7282745	.4271211	-0.54	0.589	.2307168	2.298851
	Other	6.89e-07	.0004427	-0.02	0.982	0	.
	AsianMale	3.84e+12	7.64e+15	0.01	0.988	0	.
	BlackMale	3.668586	2.369815	2.01	0.044	1.034294	13.01229
	LatinoMale	5.482559	5.61476	1.66	0.097	.7366424	40.80468
	OtherMale	8.75e+12	9.49e+15	0.03	0.978	0	.
3							
	tenure	.9958917	.011131	-0.37	0.713	.9743127	1.017949
	exper	1.016337	.0158933	1.04	0.300	.9856592	1.04797
	male	1.122801	.22105	0.59	0.556	.7633501	1.651513
	age	1.003477	.0146344	0.24	0.812	.9752	1.032574
	pyacctrati~2	1.877535	.3134683	3.77	0.000	1.353548	2.604369
	pyperstudlep	1.023528	.0177314	1.34	0.179	.9893588	1.058878
	pyperstude~s	1.01143	.0077529	1.48	0.138	.9963485	1.02674
	pyperstudw~e	1.04902	.057217	0.88	0.380	.9426633	1.167377
	pyperstudb~k	1.045019	.0609532	0.75	0.450	.932129	1.171582
	pyperstudl~o	1.033845	.0567653	0.61	0.544	.9283648	1.15131
	pyperstudm~e	1.00098	.0096578	0.10	0.919	.9822285	1.020089
	Asian	1.880547	2.721776	0.44	0.663	.1102333	32.08156
	Black	1.550524	.574222	1.18	0.236	.750316	3.204151
	Latino	.7710799	.5231378	-0.38	0.702	.2039874	2.914711
	Other	.870007	.8215866	-0.15	0.883	.1366769	5.537969
	AsianMale	.648082	1663.941	-0.00	1.000	0	.
	BlackMale	1.765749	1.190047	0.84	0.399	.4712506	6.616158
	LatinoMale	6.021919	6.44801	1.68	0.094	.7384229	49.1094
	OtherMale	1467821	1.28e+09	0.02	0.987	0	.

-----> region = 13

Iteration 0: log likelihood = -1567.4481  
Iteration 1: log likelihood = -1485.2674  
Iteration 2: log likelihood = -1483.9588  
Iteration 3: log likelihood = -1483.9037  
Iteration 4: log likelihood = -1483.8952  
Iteration 5: log likelihood = -1483.8931  
Iteration 6: log likelihood = -1483.8927  
Iteration 7: log likelihood = -1483.8926  
Iteration 8: log likelihood = -1483.8925

Multinomial logistic regression	Number of obs	=	1440
	LR chi2(38)	=	167.11
	Prob > chi2	=	0.0000
Log likelihood = -1483.8925	Pseudo R2	=	0.0533

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1					
	(base outcome)				

-----							
2	tenure	1.013811	.0107158	1.30	0.194	.9930249	1.035033
	exper	.9727827	.0112762	-2.38	0.017	.9509308	.9951367
	male	.8861459	.1404411	-0.76	0.446	.6495347	1.208949
	age	.9759603	.0097537	-2.43	0.015	.9570294	.9952656
	pyacctrati~2	1.073701	.1441204	0.53	0.596	.8253315	1.396812
	pyperstudlep	1.008188	.0080076	1.03	0.305	.9926144	1.024005
	pyperstude~s	.9859349	.0071809	-1.94	0.052	.9719607	1.00011
	pyperstudw~e	1.011489	.0195623	0.59	0.555	.9738651	1.050566
	pyperstudb~k	1.027881	.024124	1.17	0.241	.98167	1.076268
	pyperstudl~o	1.024314	.0215629	1.14	0.254	.9829117	1.067461
	pyperstudm~e	.9781744	.0114896	-1.88	0.060	.9559124	1.000955
	Asian	4.833541	5.50761	1.38	0.167	.5180339	45.0996
	Black	1.513778	.4176829	1.50	0.133	.8814509	2.599719
	Latino	1.637491	.3813402	2.12	0.034	1.03741	2.584683
	Other	1.284943	1.195036	0.27	0.787	.2076065	7.952926
	AsianMale	.1267634	.2320351	-1.13	0.259	.0035068	4.582286
	BlackMale	1.241219	.5935514	0.45	0.651	.4861875	3.168789
	LatinoMale	1.993444	.7876936	1.75	0.081	.9188794	4.324636
	OtherMale	.6301	.7211773	-0.40	0.687	.0668608	5.938096

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3	tenure	1.02761	.0102439	2.73	0.006	1.007727	1.047885
	exper	1.011108	.0123775	0.90	0.367	.9871375	1.035661
	male	.8562279	.1407541	-0.94	0.345	.6203854	1.181727
	age	.9938552	.0109085	-0.56	0.574	.9727034	1.015467
	pyacctrati~2	1.390308	.194851	2.35	0.019	1.056368	1.829812
	pyperstudlep	1.00391	.0082445	0.48	0.635	.9878799	1.020199
	pyperstude~s	1.018954	.0074972	2.55	0.011	1.004365	1.033755
	pyperstudw~e	.9845136	.0195873	-0.78	0.433	.946862	1.023662
	pyperstudb~k	.9753535	.0239555	-1.02	0.310	.9295138	1.023454
	pyperstudl~o	.9783451	.0209562	-1.02	0.307	.9381221	1.020293
	pyperstudm~e	.9649796	.0135733	-2.53	0.011	.9387397	.991953
	Asian	1.440325	2.051298	0.26	0.798	.088346	23.48194
	Black	.5257752	.1882091	-1.80	0.073	.2606753	1.060475
	Latino	1.04227	.2657838	0.16	0.871	.6322951	1.718069
	Other	2.122176	1.98037	0.81	0.420	.3407674	13.21615
	AsianMale	3.10e-06	.0021291	-0.02	0.985	0	.
	BlackMale	2.881772	1.641628	1.86	0.063	.9435495	8.801454
	LatinoMale	1.978569	.8694064	1.55	0.120	.8362255	4.681436
	OtherMale	.1796967	.2580631	-1.20	0.232	.0107678	2.998842

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -241.36858  
Iteration 1: log likelihood = -197.85605  
Iteration 2: log likelihood = -191.18248  
Iteration 3: log likelihood = -190.50499  
Iteration 4: log likelihood = -190.44554  
Iteration 5: log likelihood = -190.43364  
Iteration 6: log likelihood = -190.43111  
Iteration 7: log likelihood = -190.43053  
Iteration 8: log likelihood = -190.43039  
Iteration 9: log likelihood = -190.43036  
Iteration 10: log likelihood = -190.43036

Multinomial logistic regression	Number of obs	=	247
	LR chi2(34)	=	101.88
	Prob > chi2	=	0.0000
Log likelihood = -190.43036	Pseudo R2	=	0.2110

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.070678	.0373395	1.96	0.050	.9999387	1.146421
	exper	1.085672	.0649035	1.37	0.169	.9656333	1.220633
	male	4.856149	2.92952	2.62	0.009	1.488663	15.84119
	age	.935328	.0546259	-1.14	0.252	.8341637	1.048761
pyacctrati~2		1.015923	.491965	0.03	0.974	.3932443	2.624573
pyperstudlep		1.068568	.139114	0.51	0.610	.8279165	1.37917
pyperstude~s		.968586	.0258198	-1.20	0.231	.9192795	1.020537
pyperstudw~e		.5646567	.1702933	-1.90	0.058	.3126605	1.019755
pyperstudb~k		.5199205	.1797915	-1.89	0.059	.2639894	1.02397
pyperstudl~o		.5951166	.1742937	-1.77	0.076	.3352031	1.056565
pyperstudm~e		1.046326	.0672595	0.70	0.481	.9224665	1.186817
	Asian	(omitted)					
	Black	19.66539	31.40267	1.87	0.062	.859927	449.7214
	Latino	4.07e-06	.0071566	-0.01	0.994	0	.
	Other	20.42637	104316.7	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.1090687	381.6365	-0.00	0.999	0	.
	LatinoMale	2418775	4.26e+09	0.01	0.993	0	.
	OtherMale	.116893	596.9681	-0.00	1.000	0	.
3							
	tenure	1.047665	.0241766	2.02	0.044	1.001336	1.096139
	exper	1.065971	.0340787	2.00	0.046	1.001227	1.134901
	male	2.704681	.94244	2.86	0.004	1.366208	5.354456
	age	.9412356	.0281222	-2.03	0.043	.8877	.9979999
pyacctrati~2		2.2932	.6520813	2.92	0.004	1.313407	4.003914
pyperstudlep		1.01902	.0715391	0.27	0.788	.8880249	1.169339
pyperstude~s		1.055076	.0163491	3.46	0.001	1.023514	1.087611
pyperstudw~e		.8711193	.1672395	-0.72	0.472	.597946	1.269092
pyperstudb~k		.7912739	.1699115	-1.09	0.276	.5194539	1.205332
pyperstudl~o		.8774204	.1655341	-0.69	0.488	.6062057	1.269976
pyperstudm~e		.9479044	.0326725	-1.55	0.121	.8859826	1.014154
	Asian	(omitted)					
	Black	4.423785	7.023071	0.94	0.349	.1969933	99.34284
	Latino	1.943642	2.515055	0.51	0.608	.1538754	24.55067
	Other	8017013	1.86e+10	0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	2197206	3.40e+09	0.01	0.992	0	.
	LatinoMale	.604234	.9805993	-0.31	0.756	.0251077	14.5413
	OtherMale	3.69e-15	1.21e-11	-0.01	0.992	0	.

-----> region = 15

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -272.22599  
Iteration 1: log likelihood = -242.22037  
Iteration 2: log likelihood = -238.97322  
Iteration 3: log likelihood = -238.86576  
Iteration 4: log likelihood = -238.85383  
Iteration 5: log likelihood = -238.85165  
Iteration 6: log likelihood = -238.85116  
Iteration 7: log likelihood = -238.85105  
Iteration 8: log likelihood = -238.85102  
Iteration 9: log likelihood = -238.85102

Multinomial logistic regression

Number of obs = 267  
LR chi2(32) = 66.75  
Prob > chi2 = 0.0003



Log likelihood = -238.85102

Pseudo R2 = 0.1226

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.061259	.0434523	1.45	0.146	.9794213	1.149934
exper	.9360289	.0398068	-1.55	0.120	.861172	1.017393
male	.851044	.4450928	-0.31	0.758	.3053351	2.372069
age	1.029176	.0296879	1.00	0.319	.9726035	1.08904
pyacctrati~2	.3860036	.1946168	-1.89	0.059	.1436909	1.036939
pyperstudlep	.9621451	.0390674	-0.95	0.342	.888542	1.041845
pyperstude~s	.997162	.016501	-0.17	0.864	.9653393	1.030034
pyperstudw~e	1.002962	.3643435	0.01	0.994	.4921193	2.044082
pyperstudb~k	1.070582	.410839	0.18	0.859	.5046252	2.271282
pyperstudl~o	1.006151	.3668288	0.02	0.987	.4924102	2.055889
pyperstudm~e	.9964822	.0227032	-0.15	0.877	.9529638	1.041988
Asian	(omitted)					
Black	9.15e+07	2.65e+11	0.01	0.995	0	.
Latino	4.264706	2.579189	2.40	0.016	1.303477	13.95323
Other	6.69e-07	.0011781	-0.01	0.994	0	.
AsianMale	(omitted)					
BlackMale	8.43e-08	.0002443	-0.01	0.996	0	.
LatinoMale	1.927676	1.770058	0.71	0.475	.3187365	11.65833
OtherMale	(omitted)					
3						
tenure	.9633874	.0209295	-1.72	0.086	.9232274	1.005294
exper	1.078063	.0286294	2.83	0.005	1.023386	1.135662
male	1.017505	.3308018	0.05	0.957	.538024	1.924293
age	.9819011	.0230517	-0.78	0.437	.9377443	1.028137
pyacctrati~2	.9225218	.2639241	-0.28	0.778	.5265705	1.616206
pyperstudlep	1.014158	.0284468	0.50	0.616	.9599079	1.071474
pyperstude~s	1.011598	.0103377	1.13	0.259	.9915384	1.032064
pyperstudw~e	.9661002	.2154844	-0.15	0.877	.6239725	1.495819
pyperstudb~k	.9519336	.2235148	-0.21	0.834	.6008193	1.508237
pyperstudl~o	.9579778	.2141438	-0.19	0.848	.6181304	1.484673
pyperstudm~e	.9701076	.0182383	-1.61	0.106	.9350119	1.006521
Asian	(omitted)					
Black	1.501798	6454.951	0.00	1.000	0	.
Latino	.5285987	.3038417	-1.11	0.267	.1713367	1.630804
Other	1.28e-06	.0014619	-0.01	0.991	0	.
AsianMale	(omitted)					
BlackMale	1.157898	4976.82	0.00	1.000	0	.
LatinoMale	2.105992	1.869134	0.84	0.401	.3698183	11.99292
OtherMale	(omitted)					

-----&gt; region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -513.59118  
Iteration 1: log likelihood = -466.4979  
Iteration 2: log likelihood = -464.19248  
Iteration 3: log likelihood = -463.84262  
Iteration 4: log likelihood = -463.7714  
Iteration 5: log likelihood = -463.75557  
Iteration 6: log likelihood = -463.75298  
Iteration 7: log likelihood = -463.75236  
Iteration 8: log likelihood = -463.75222  
Iteration 9: log likelihood = -463.75219

Multinomial logistic regression

Number of obs = 472

Log likelihood = -463.75219

LR chi2(34)	=	99.68
Prob > chi2	=	0.0000
Pseudo R2	=	0.0970

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.031737	.021967	1.47	0.142	.9895689	1.075703
	exper	.9478403	.025268	-2.01	0.044	.8995875	.9986814
	male	1.134856	.3162805	0.45	0.650	.6572246	1.959601
	age	.9731323	.021876	-1.21	0.226	.9311868	1.016967
pyacctrati~2		1.143458	.2673739	0.57	0.566	.7230766	1.80824
pyperstudlep		1.036433	.0171865	2.16	0.031	1.00329	1.070672
pyperstude~s		.9821171	.0088556	-2.00	0.045	.9649129	.9996279
pyperstudw~e		.9641689	.0336299	-1.05	0.295	.9004579	1.032388
pyperstudb~k		1.007619	.0398368	0.19	0.848	.932489	1.088803
pyperstudl~o		.967994	.0326294	-0.97	0.335	.9061084	1.034106
pyperstudm~e		1.013277	.012618	1.06	0.290	.9888452	1.038312
	Asian	(omitted)					
	Black	.617462	.6339958	-0.47	0.639	.0825308	4.6196
	Latino	2.958914	1.836919	1.75	0.081	.8763937	9.99
	Other	3.015285	4.210473	0.79	0.429	.1953128	46.5507
	AsianMale	(omitted)					
	BlackMale	2.20e-06	.0022273	-0.01	0.990	0	.
	LatinoMale	481234.8	2.41e+08	0.03	0.979	0	.
	OtherMale	4823034	8.04e+09	0.01	0.993	0	.
3							
	tenure	1.019108	.0165929	1.16	0.245	.9871003	1.052154
	exper	1.001089	.0216316	0.05	0.960	.9595776	1.044397
	male	2.472245	.6186163	3.62	0.000	1.51391	4.037225
	age	.9937167	.0204025	-0.31	0.759	.9545224	1.03452
pyacctrati~2		1.770211	.3596341	2.81	0.005	1.188764	2.636053
pyperstudlep		1.032728	.016434	2.02	0.043	1.001015	1.065446
pyperstude~s		1.007765	.0074969	1.04	0.298	.993178	1.022566
pyperstudw~e		1.047101	.0475736	1.01	0.311	.9578899	1.144621
pyperstudb~k		1.045051	.0523242	0.88	0.379	.9473691	1.152806
pyperstudl~o		1.029919	.0458647	0.66	0.508	.9438368	1.123851
pyperstudm~e		.9865702	.0147461	-0.90	0.366	.9580876	1.0159
	Asian	(omitted)					
	Black	1.28253	1.266325	0.25	0.801	.1851914	8.882066
	Latino	2.062351	1.380192	1.08	0.279	.5555341	7.65622
	Other	2.04e-06	.0019615	-0.01	0.989	0	.
	AsianMale	(omitted)					
	BlackMale	.6014212	1.060429	-0.29	0.773	.0189817	19.05558
	LatinoMale	466717.8	2.34e+08	0.03	0.979	0	.
	OtherMale	321068.6	7.45e+08	0.01	0.996	0	.

-----> region = 17

note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -435.73192  
Iteration 1: log likelihood = -386.89918  
Iteration 2: log likelihood = -385.51949  
Iteration 3: log likelihood = -385.45925  
Iteration 4: log likelihood = -385.44987  
Iteration 5: log likelihood = -385.44764  
Iteration 6: log likelihood = -385.44715  
Iteration 7: log likelihood = -385.44704  
Iteration 8: log likelihood = -385.44702

Multinomial logistic regression

Number of obs = 437  
 LR chi2(32) = 100.57  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.1154

Log likelihood = -385.44702

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.9945626	.0240835	-0.23	0.822	.9484623	1.042904
exper	1.06885	.0372632	1.91	0.056	.9982549	1.144437
male	1.933782	.6628193	1.92	0.054	.9877597	3.785852
age	.928998	.0291752	-2.35	0.019	.8735399	.9879769
pyacctrati~2	.9839576	.3065547	-0.05	0.959	.5342964	1.812052
pyperstudlep	.984068	.0381847	-0.41	0.679	.9120026	1.061828
pyperstude~s	.9762399	.0141101	-1.66	0.096	.9489726	1.004291
pyperstudw~e	1.130165	.1168762	1.18	0.237	.9228147	1.384104
pyperstudb~k	1.133828	.1258769	1.13	0.258	.9121099	1.409441
pyperstudl~o	1.136452	.1175291	1.24	0.216	.927944	1.391811
pyperstudm~e	1.010266	.0211143	0.49	0.625	.9697185	1.052508
Asian	1.97e-06	.0033653	-0.01	0.994	0	.
Black	.0000101	.0172017	-0.01	0.995	0	.
Latino	5.279164	3.55841	2.47	0.014	1.408687	19.78408
Other	(omitted)					
AsianMale	(omitted)					
BlackMale	2.11e+13	1.45e+17	0.00	0.996	0	.
LatinoMale	1.122549	1.130944	0.11	0.909	.1558253	8.086722
OtherMale	(omitted)					
3						
tenure	1.01425	.0175055	0.82	0.412	.9805139	1.049147
exper	1.05658	.0242739	2.40	0.017	1.01006	1.105244
male	4.411768	1.174771	5.57	0.000	2.617904	7.434841
age	.9778645	.0200493	-1.09	0.275	.9393476	1.017961
pyacctrati~2	1.569086	.3531521	2.00	0.045	1.00941	2.439077
pyperstudlep	1.017039	.0250509	0.69	0.493	.9691065	1.067342
pyperstude~s	1.022784	.0107927	2.13	0.033	1.001848	1.044157
pyperstudw~e	.994759	.0731561	-0.07	0.943	.8612301	1.148991
pyperstudb~k	.9857183	.0764255	-0.19	0.853	.8467531	1.14749
pyperstudl~o	.9896371	.0727409	-0.14	0.887	.8568611	1.142987
pyperstudm~e	.9792666	.0173487	-1.18	0.237	.9458472	1.013867
Asian	6.90e-06	.0076924	-0.01	0.991	0	.
Black	6.75e-06	.0075316	-0.01	0.991	0	.
Latino	2.911902	1.836173	1.69	0.090	.8461166	10.02128
Other	(omitted)					
AsianMale	(omitted)					
BlackMale	74486.68	7.77e+08	0.00	0.999	0	.
LatinoMale	.8406666	.7893126	-0.18	0.853	.1334811	5.294534
OtherMale	(omitted)					

-----> region = 18

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -424.8665  
 Iteration 1: log likelihood = -396.52911  
 Iteration 2: log likelihood = -395.82335  
 Iteration 3: log likelihood = -395.78346  
 Iteration 4: log likelihood = -395.77809  
 Iteration 5: log likelihood = -395.77742  
 Iteration 6: log likelihood = -395.77729  
 Iteration 7: log likelihood = -395.77726

Multinomial logistic regression

Number of obs = 401  
 LR chi2(34) = 58.18  
 Prob > chi2 = 0.0061  
 Pseudo R2 = 0.0685

Log likelihood = -395.77726

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.013363	.0215806	0.62	0.533	.9719363	1.056555
exper	.9780496	.025445	-0.85	0.394	.9294285	1.029214
male	1.20008	.4147236	0.53	0.598	.6096098	2.36248
age	.9794989	.0210305	-0.96	0.335	.9391351	1.021598
pyacctrati~2	.6320902	.2292875	-1.26	0.206	.3104632	1.28691
pyperstudlep	1.007032	.0191223	0.37	0.712	.9702423	1.045218
pyperstude~s	.9967653	.014077	-0.23	0.819	.9695533	1.024741
pyperstudw~e	.8314829	.1272245	-1.21	0.228	.6160439	1.122264
pyperstudb~k	.8612263	.1369995	-0.94	0.348	.6305397	1.176311
pyperstudl~o	.825183	.1256451	-1.26	0.207	.6122707	1.112134
pyperstudm~e	1.017686	.017779	1.00	0.316	.9834299	1.053136
Asian	(omitted)					
Black	2.036684	2.036408	0.71	0.477	.2869707	14.45473
Latino	1.998974	.8831961	1.57	0.117	.840863	4.752138
Other	.4799999	455.3744	-0.00	0.999	0	.
AsianMale	(omitted)					
BlackMale	2.704431	4.241678	0.63	0.526	.1250364	58.49455
LatinoMale	.5658496	.338973	-0.95	0.342	.174899	1.830689
OtherMale	4052402	5.04e+09	0.01	0.990	0	.
3						
tenure	1.005148	.0170937	0.30	0.763	.9721969	1.039215
exper	1.066169	.0263773	2.59	0.010	1.015704	1.119142
male	1.488074	.4403275	1.34	0.179	.8332018	2.657656
age	.9479647	.020745	-2.44	0.015	.9081648	.9895088
pyacctrati~2	1.59783	.4548189	1.65	0.100	.9146143	2.791408
pyperstudlep	.9998142	.016943	-0.01	0.991	.967152	1.03358
pyperstude~s	1.025671	.0131283	1.98	0.048	1.00026	1.051728
pyperstudw~e	1.023669	.1436742	0.17	0.868	.7774844	1.347807
pyperstudb~k	1.007858	.1485382	0.05	0.958	.755004	1.345394
pyperstudl~o	1.00533	.1407005	0.04	0.970	.7641509	1.32263
pyperstudm~e	1.015888	.0162286	0.99	0.324	.9845734	1.048199
Asian	(omitted)					
Black	3.012573	2.967747	1.12	0.263	.4369195	20.77177
Latino	.9906845	.4043742	-0.02	0.982	.4451382	2.204834
Other	383846	2.11e+08	0.02	0.981	0	.
AsianMale	(omitted)					
BlackMale	.7296121	1.162467	-0.20	0.843	.0321292	16.56853
LatinoMale	.4867575	.2794903	-1.25	0.210	.1579657	1.4999
OtherMale	2.63e-06	.0036134	-0.01	0.993	0	.

-----> region = 19

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -810.58857  
 Iteration 1: log likelihood = -730.73965  
 Iteration 2: log likelihood = -726.05995  
 Iteration 3: log likelihood = -725.90045  
 Iteration 4: log likelihood = -725.86623  
 Iteration 5: log likelihood = -725.85792  
 Iteration 6: log likelihood = -725.8563  
 Iteration 7: log likelihood = -725.85604  
 Iteration 8: log likelihood = -725.85598

Iteration 9: log likelihood = -725.85596

Multinomial logistic regression	Number of obs	=	750
	LR chi2(36)	=	169.47
	Prob > chi2	=	0.0000
Log likelihood = -725.85596	Pseudo R2	=	0.1045

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	tenure	.9834978	.014126	-1.16	0.247	.9561974	1.011578
	exper	.9824797	.0172364	-1.01	0.314	.9492712	1.01685
	male	1.027792	.3170006	0.09	0.929	.5615263	1.881223
	age	1.028634	.0138778	2.09	0.036	1.001791	1.056197
	pyacctrati~2	1.19726	.2431152	0.89	0.375	.8041627	1.782515
	pyperstudlep	.9922516	.0059966	-1.29	0.198	.9805678	1.004075
	pyperstude~s	1.007836	.0100688	0.78	0.435	.988294	1.027765
	pyperstudw~e	1.184389	.1257403	1.59	0.111	.9618935	1.458351
	pyperstudb~k	1.092978	.1206831	0.81	0.421	.8802877	1.357058
	pyperstudl~o	1.146451	.1146567	1.37	0.172	.9423811	1.394711
	pyperstudm~e	1.005151	.0113593	0.45	0.649	.9831315	1.027663
	Asian	1.351833	.6275.05	0.00	1.000	0	.
	Black	1.007763	.586165	0.01	0.989	.322298	3.15108
	Latino	.686891	.1663039	-1.55	0.121	.4273685	1.10401
	Other	.4113753	.5149415	-0.71	0.478	.0353787	4.783379
	AsianMale	(omitted)					
	BlackMale	.3178113	.3426728	-1.06	0.288	.0384043	2.630015
	LatinoMale	.9172961	.3473553	-0.23	0.820	.4366992	1.9268
	OtherMale	1407641	1.42e+09	0.01	0.989	0	.
2		(base outcome)					
3							
	tenure	1.015047	.0152795	0.99	0.321	.9855374	1.045441
	exper	1.115263	.0252307	4.82	0.000	1.066892	1.165827
	male	1.592802	.5757344	1.29	0.198	.7843074	3.234724
	age	.976361	.0190897	-1.22	0.221	.9396537	1.014502
	pyacctrati~2	1.585174	.4007932	1.82	0.068	.9657405	2.601918
	pyperstudlep	1.017581	.0071629	2.48	0.013	1.003639	1.031718
	pyperstude~s	1.030137	.0130858	2.34	0.019	1.004806	1.056107
	pyperstudw~e	1.093675	.1121991	0.87	0.383	.8944668	1.337249
	pyperstudb~k	1.035115	.1131914	0.32	0.752	.8354269	1.282534
	pyperstudl~o	1.041954	.0999779	0.43	0.668	.8633235	1.257544
	pyperstudm~e	.982052	.0152595	-1.17	0.244	.9525947	1.01242
	Asian	854207.3	2.79e+09	0.00	0.997	0	.
	Black	.8741149	.6972111	-0.17	0.866	.1830769	4.173529
	Latino	.7599927	.2215035	-0.94	0.346	.4292622	1.345539
	Other	9.57e-07	.0007405	-0.02	0.986	0	.
	AsianMale	(omitted)					
	BlackMale	.3470279	.5021478	-0.73	0.465	.0203554	5.916287
	LatinoMale	.7303811	.3270004	-0.70	0.483	.3037081	1.756478
	OtherMale	5.94e+12	7.55e+15	0.02	0.982	0	.

-----> region = 20

Iteration 0: log likelihood = -2091.9463  
Iteration 1: log likelihood = -1990.9741  
Iteration 2: log likelihood = -1986.7222  
Iteration 3: log likelihood = -1986.4615  
Iteration 4: log likelihood = -1986.4159  
Iteration 5: log likelihood = -1986.4058  
Iteration 6: log likelihood = -1986.4033  
Iteration 7: log likelihood = -1986.4028  
Iteration 8: log likelihood = -1986.4027

Iteration 9: log likelihood = -1986.4027  
 Iteration 10: log likelihood = -1986.4027

Multinomial logistic regression	Number of obs	=	2037
	LR chi2(38)	=	211.09
	Prob > chi2	=	0.0000
Log likelihood = -1986.4027	Pseudo R2	=	0.0505

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9965557	.0082182	-0.42	0.676	.9805778	1.012794
	exper	1.003885	.0101953	0.38	0.703	.9841001	1.024067
	male	1.419689	.2152066	2.31	0.021	1.054781	1.910841
	age	.9764823	.0082642	-2.81	0.005	.9604183	.992815
pyacctrati~2		.9190313	.1145692	-0.68	0.498	.7198086	1.173393
pyperstudlep		.9836338	.0068432	-2.37	0.018	.9703124	.9971382
pyperstude~s		1.000682	.0055913	0.12	0.903	.9897832	1.011701
pyperstudw~e		1.016054	.0383774	0.42	0.673	.9435525	1.094126
pyperstudb~k		1.009087	.0408924	0.22	0.823	.9320394	1.092503
pyperstudl~o		1.014585	.0391594	0.38	0.708	.940665	1.094313
pyperstudm~e		.993945	.0073778	-0.82	0.413	.9795896	1.008511
	Asian	.7142264	.8303102	-0.29	0.772	.0731624	6.972426
	Black	1.779192	.5016852	2.04	0.041	1.02378	3.091998
	Latino	1.142425	.176361	0.86	0.388	.8441591	1.546078
	Other	.8930393	.7557814	-0.13	0.894	.1700187	4.690774
	AsianMale	1.979753	3.641738	0.37	0.710	.0538063	72.84316
	BlackMale	.9468293	.4043411	-0.13	0.898	.4099852	2.18663
	LatinoMale	.9727349	.2260085	-0.12	0.905	.6169123	1.533789
	OtherMale	1.049801	1.317276	0.04	0.969	.0897493	12.27957
3							
	tenure	1.015492	.0083713	1.86	0.062	.9992162	1.032032
	exper	1.087873	.0145292	6.31	0.000	1.059765	1.116725
	male	1.474098	.2520934	2.27	0.023	1.054284	2.06108
	age	.9485952	.0113548	-4.41	0.000	.9265992	.9711133
pyacctrati~2		1.269302	.1695792	1.78	0.074	.9768868	1.649247
pyperstudlep		1.013309	.006633	2.02	0.043	1.000392	1.026393
pyperstude~s		1.029936	.0064783	4.69	0.000	1.017317	1.042712
pyperstudw~e		.9918632	.039611	-0.20	0.838	.9171878	1.072619
pyperstudb~k		.9628346	.0413384	-0.88	0.378	.8851282	1.047363
pyperstudl~o		.9646083	.039337	-0.88	0.377	.8905099	1.044872
pyperstudm~e		.9839129	.0090408	-1.77	0.078	.9663519	1.001793
	Asian	3.85e-07	.0005725	-0.01	0.992	0	.
	Black	1.696082	.5252392	1.71	0.088	.9243756	3.11204
	Latino	.861928	.1466093	-0.87	0.382	.6175701	1.202973
	Other	3.91e-07	.0004595	-0.01	0.990	0	.
	AsianMale	1.135279	3070.664	0.00	1.000	0	.
	BlackMale	.5554956	.3076348	-1.06	0.288	.1876217	1.644668
	LatinoMale	.9164345	.2451502	-0.33	0.744	.542502	1.548109
	OtherMale	1800987	2.12e+09	0.01	0.990	0	.

by pyschooltype:

```
. bysort pyschooltype : mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2006a==1, rrr iter(20)
```

```
-----> pyschooltype =
no observations
```

-----> pyschooltype = B

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -381.34099  
 Iteration 1: log likelihood = -347.85883  
 Iteration 2: log likelihood = -340.26665  
 Iteration 3: log likelihood = -335.67685  
 Iteration 4: log likelihood = -335.4932  
 Iteration 5: log likelihood = -335.4823  
 Iteration 6: log likelihood = -335.48154  
 Iteration 7: log likelihood = -335.48138  
 Iteration 8: log likelihood = -335.48134  
 Iteration 9: log likelihood = -335.48133

Multinomial logistic regression	Number of obs	=	418
	LR chi2(34)	=	91.72
	Prob > chi2	=	0.0000
Log likelihood = -335.48133	Pseudo R2	=	0.1203

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.076232	.0334019	2.37	0.018	1.012717	1.143731
	exper	.942772	.0349444	-1.59	0.112	.8767108	1.013811
	male	2.07479	.93627	1.62	0.106	.8567634	5.024438
	age	.9765877	.0291303	-0.79	0.427	.9211302	1.035384
pyacctrati~2		.6817999	.2786596	-0.94	0.349	.3060279	1.518983
pyperstudlep		.9929996	.0263021	-0.27	0.791	.9427637	1.045912
pyperstude~s		.9915335	.0151265	-0.56	0.577	.9623249	1.021629
pyperstudw~e		.9099342	.0308659	-2.78	0.005	.8514052	.9724866
pyperstudb~k		.933693	.0345138	-1.86	0.063	.8684396	1.00385
pyperstudl~o		.9397404	.0337901	-1.73	0.084	.8757928	1.008357
pyperstudm~e		.9473156	.0288729	-1.78	0.076	.8923829	1.00563
	Asian	(omitted)					
	Black	3.855862	3.432607	1.52	0.130	.673528	22.07432
	Latino	1.577541	1.378824	0.52	0.602	.2844471	8.749025
	Other	12.17352	22.211	1.37	0.171	.3406979	434.9732
	AsianMale	(omitted)					
	BlackMale	.5746528	.675532	-0.47	0.637	.0573822	5.754848
	LatinoMale	6.465887	7.599413	1.59	0.112	.6459569	64.72211
	OtherMale	3.99e-07	.000569	-0.01	0.992	0	.
3							
	tenure	.9807477	.0159591	-1.19	0.232	.949962	1.012531
	exper	1.016549	.0187526	0.89	0.374	.9804514	1.053976
	male	1.423456	.3438962	1.46	0.144	.8865438	2.285535
	age	.9972953	.018055	-0.15	0.881	.9625287	1.033318
pyacctrati~2		.6353879	.1561053	-1.85	0.065	.3925649	1.02841
pyperstudlep		.9965235	.0187605	-0.18	0.853	.9604237	1.03398
pyperstude~s		1.004079	.0080833	0.51	0.613	.9883604	1.020048
pyperstudw~e		.9337177	.0254466	-2.52	0.012	.8851518	.9849483
pyperstudb~k		.9211411	.0270581	-2.80	0.005	.869606	.9757304
pyperstudl~o		.9313957	.0262243	-2.52	0.012	.8813895	.9842391
pyperstudm~e		1.00815	.0093512	0.88	0.382	.989988	1.026646
	Asian	(omitted)					
	Black	2.183807	1.79791	0.95	0.343	.4349377	10.96482
	Latino	.220333	.2399938	-1.39	0.165	.0260569	1.863098
	Other	5.98e-07	.0012834	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	.5191732	.5549438	-0.61	0.540	.0638945	4.218528
	LatinoMale	24.45608	33.80431	2.31	0.021	1.628612	367.2453
	OtherMale	1.773583	4093.171	0.00	1.000	0	.

```
-----> pyschooltype = E
```

```
Iteration 0: log likelihood = -11168.659
Iteration 1: log likelihood = -10550.695
Iteration 2: log likelihood = -10544.723
Iteration 3: log likelihood = -10544.717
Iteration 4: log likelihood = -10544.717
```

```
Multinomial logistic regression      Number of obs   =      10279
                                      LR chi2(38)        =      1247.88
                                      Prob > chi2         =      0.0000
Log likelihood = -10544.717          Pseudo R2       =      0.0559
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.015546	.0042739	3.67	0.000	1.007204	1.023957
	exper	1.006362	.005403	1.18	0.237	.9958283	1.017008
	male	1.537765	.1332481	4.97	0.000	1.297576	1.822413
	age	.9777142	.0041106	-5.36	0.000	.9696906	.9858042
pyacctrati~2		.9538169	.0434054	-1.04	0.299	.8724275	1.042799
pyperstudlep		1.009361	.0018642	5.04	0.000	1.005714	1.013021
pyperstude~s		.9890224	.0021893	-4.99	0.000	.9847407	.9933228
pyperstudw~e		1.000781	.0054749	0.14	0.887	.9901074	1.011569
pyperstudb~k		1.001466	.0061986	0.24	0.813	.9893899	1.013689
pyperstudl~o		1.002458	.005811	0.42	0.672	.9911335	1.013913
pyperstudm~e		1.010139	.0037814	2.69	0.007	1.002755	1.017578
	Asian	1.057347	.4653438	0.13	0.899	.4462716	2.505161
	Black	1.13998	.1074238	1.39	0.164	.947733	1.371225
	Latino	1.184323	.0936338	2.14	0.032	1.014316	1.382824
	Other	.9489309	.2774081	-0.18	0.858	.5350525	1.682956
	AsianMale	1.332301	1.077131	0.35	0.723	.2731667	6.49796
	BlackMale	.8754224	.1732364	-0.67	0.501	.5939835	1.290212
	LatinoMale	.7854689	.1153915	-1.64	0.100	.5889533	1.047556
	OtherMale	1.469941	.8309076	0.68	0.496	.4854502	4.450975
3							
	tenure	1.020691	.0036543	5.72	0.000	1.013554	1.027879
	exper	1.071268	.0054737	13.47	0.000	1.060594	1.08205
	male	2.542572	.1966065	12.07	0.000	2.18501	2.958647
	age	.9808826	.0042193	-4.49	0.000	.9726477	.9891872
pyacctrati~2		.9812732	.0429094	-0.43	0.666	.9006755	1.069083
pyperstudlep		.9988038	.0018142	-0.66	0.510	.9952543	1.002366
pyperstude~s		1.005145	.0020899	2.47	0.014	1.001057	1.009249
pyperstudw~e		1.007316	.0056	1.31	0.190	.9963995	1.018351
pyperstudb~k		.9953309	.0062361	-0.75	0.455	.9831832	1.007629
pyperstudl~o		.9965105	.0058566	-0.59	0.552	.9850975	1.008056
pyperstudm~e		1.000815	.0037282	0.22	0.827	.9935339	1.008148
	Asian	1.506146	.6641636	0.93	0.353	.6346198	3.574543
	Black	.9371899	.0907305	-0.67	0.503	.7752142	1.133009
	Latino	1.250589	.0982335	2.85	0.004	1.072144	1.458735
	Other	.7586407	.2314694	-0.91	0.365	.4171818	1.37958
	AsianMale	.6227451	.5207249	-0.57	0.571	.120936	3.206751
	BlackMale	.5551575	.1111993	-2.94	0.003	.3749028	.8220794
	LatinoMale	.5600793	.0793846	-4.09	0.000	.4242304	.7394304
	OtherMale	.3532953	.2529601	-1.45	0.146	.0868312	1.437473

```
-----> pyschooltype = M
```



```

Iteration 0: log likelihood = -6029.9181
Iteration 1: log likelihood = -5732.1575
Iteration 2: log likelihood = -5726.7195
Iteration 3: log likelihood = -5726.4794
Iteration 4: log likelihood = -5726.4248
Iteration 5: log likelihood = -5726.4138
Iteration 6: log likelihood = -5726.4117
Iteration 7: log likelihood = -5726.4115
Iteration 8: log likelihood = -5726.4115
Iteration 9: log likelihood = -5726.4115

```

```

Multinomial logistic regression      Number of obs   =      5709
                                     LR chi2(38)        =      607.01
                                     Prob > chi2         =      0.0000
Log likelihood = -5726.4115          Pseudo R2        =      0.0503

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.007989	.0050088	1.60	0.109	.9982198	1.017854
	exper	1.009164	.0062431	1.47	0.140	.9970014	1.021474
	male	1.546415	.1231131	5.48	0.000	1.323002	1.807555
	age	.9688566	.0048738	-6.29	0.000	.9593512	.9784563
pyacctrati~2		.9785203	.0679396	-0.31	0.754	.854024	1.121165
pyperstudlep		1.011215	.0037745	2.99	0.003	1.003844	1.01864
pyperstude~s		.9925757	.0027885	-2.65	0.008	.9871255	.9980561
pyperstudw~e		1.004722	.0075465	0.63	0.531	.9900393	1.019622
pyperstudb~k		1.010147	.0085896	1.19	0.235	.993451	1.027123
pyperstudl~o		1.004368	.0080957	0.54	0.589	.9886251	1.020361
pyperstudm~e		1.009614	.0043512	2.22	0.026	1.001122	1.018178
	Asian	1.336333	.5305723	0.73	0.465	.6137016	2.909862
	Black	.8950533	.1072271	-0.93	0.355	.7077425	1.131938
	Latino	2.129769	.2525029	6.38	0.000	1.688166	2.686891
	Other	.7556312	.2913029	-0.73	0.467	.3549472	1.60863
	AsianMale	1.215337	.8686598	0.27	0.785	.2994345	4.932776
	BlackMale	1.705277	.3094005	2.94	0.003	1.194966	2.433517
	LatinoMale	.8437765	.1349799	-1.06	0.288	.6166797	1.154503
	OtherMale	2.322543	1.336543	1.46	0.143	.7518422	7.174652
3							
	tenure	1.009298	.0051243	1.82	0.068	.9993046	1.019392
	exper	1.071923	.0079692	9.34	0.000	1.056417	1.087657
	male	1.878492	.1667692	7.10	0.000	1.578487	2.235514
	age	.962267	.0063077	-5.87	0.000	.9499833	.9747095
pyacctrati~2		.9629765	.0772042	-0.47	0.638	.8229488	1.12683
pyperstudlep		1.003362	.0047798	0.70	0.481	.994037	1.012774
pyperstude~s		1.008341	.0032723	2.56	0.010	1.001947	1.014775
pyperstudw~e		1.028352	.0102843	2.80	0.005	1.008392	1.048708
pyperstudb~k		1.012318	.0113839	1.09	0.276	.9902503	1.034878
pyperstudl~o		1.011841	.0107726	1.11	0.269	.9909458	1.033177
pyperstudm~e		.9975898	.0058607	-0.41	0.681	.9861689	1.009143
	Asian	.4386588	.3345783	-1.08	0.280	.0983761	1.955978
	Black	.8038998	.1263141	-1.39	0.165	.5908203	1.093826
	Latino	1.611366	.2421433	3.17	0.001	1.20028	2.163247
	Other	.4538059	.2829075	-1.27	0.205	.1337279	1.539991
	AsianMale	2.24e-06	.0014741	-0.02	0.984	0	.
	BlackMale	2.242862	.4868767	3.72	0.000	1.465632	3.432258
	LatinoMale	.8921258	.172914	-0.59	0.556	.6101606	1.304392
	OtherMale	3.298289	2.589257	1.52	0.128	.7080593	15.36412

```

-----> pyschooltype = S

```

```

Iteration 0: log likelihood = -6954.568
Iteration 1: log likelihood = -6646.4779
Iteration 2: log likelihood = -6634.6483
Iteration 3: log likelihood = -6634.3924
Iteration 4: log likelihood = -6634.3369
Iteration 5: log likelihood = -6634.3233
Iteration 6: log likelihood = -6634.3206
Iteration 7: log likelihood = -6634.32
Iteration 8: log likelihood = -6634.3199
Iteration 9: log likelihood = -6634.3198

```

```

Multinomial logistic regression      Number of obs   =      7180
                                     LR chi2(38)        =      640.50
                                     Prob > chi2         =      0.0000
Log likelihood = -6634.3198          Pseudo R2        =      0.0460

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.005076	.003897	1.31	0.192	.9974669	1.012743
	exper	1.00868	.0050612	1.72	0.085	.9988093	1.018649
	male	1.12005	.0745624	1.70	0.089	.9830428	1.276152
	age	.9703144	.0042885	-6.82	0.000	.9619453	.9787562
pyacctrati~2		.9676917	.0766646	-0.41	0.678	.8285167	1.130245
pyperstudlep		1.020022	.004751	4.26	0.000	1.010752	1.029376
pyperstude~s		.9862338	.0023284	-5.87	0.000	.9816807	.9908081
pyperstudw~e		.9997855	.0061978	-0.03	0.972	.9877114	1.012007
pyperstudb~k		1.00601	.0070141	0.86	0.390	.992356	1.019852
pyperstudl~o		1.005306	.0067044	0.79	0.427	.9922514	1.018533
pyperstudm~e		1.004205	.0035759	1.18	0.239	.9972207	1.011238
	Asian	.6220012	.3613343	-0.82	0.414	.1992092	1.942107
	Black	1.32932	.1523855	2.48	0.013	1.061824	1.664203
	Latino	1.735364	.2030839	4.71	0.000	1.379676	2.182749
	Other	.5740975	.251812	-1.27	0.206	.2430131	1.356256
	AsianMale	2.352111	1.845711	1.09	0.276	.5052617	10.94962
	BlackMale	1.990454	.323689	4.23	0.000	1.447202	2.737632
	LatinoMale	1.142452	.1659789	0.92	0.359	.8593556	1.518809
	OtherMale	2.190614	1.221644	1.41	0.160	.7343021	6.535173
3							
	tenure	.9974427	.0046538	-0.55	0.583	.9883631	1.006606
	exper	1.062559	.0077258	8.35	0.000	1.047524	1.07781
	male	1.849656	.16786	6.78	0.000	1.548256	2.20973
	age	.9634118	.0066072	-5.44	0.000	.9505487	.976449
pyacctrati~2		1.211266	.1244662	1.87	0.062	.9903131	1.481517
pyperstudlep		.9996439	.0070487	-0.05	0.960	.9859237	1.013555
pyperstude~s		1.017035	.0030954	5.55	0.000	1.010986	1.02312
pyperstudw~e		1.040162	.0116378	3.52	0.000	1.0176	1.063223
pyperstudb~k		1.025767	.0125727	2.08	0.038	1.001419	1.050708
pyperstudl~o		1.024398	.012015	2.06	0.040	1.001118	1.04822
pyperstudm~e		.992551	.0051074	-1.45	0.146	.9825909	1.002612
	Asian	.5378449	.5637573	-0.59	0.554	.068936	4.196314
	Black	.5927777	.1374002	-2.26	0.024	.3763496	.9336675
	Latino	1.197068	.2260859	0.95	0.341	.8267147	1.733334
	Other	.5526064	.4123965	-0.79	0.427	.1279927	2.385869
	AsianMale	7.65e-06	.0027737	-0.03	0.974	0	3.1e+303
	BlackMale	2.749637	.7681753	3.62	0.000	1.590275	4.75421
	LatinoMale	.9791882	.2150281	-0.10	0.924	.6367131	1.505874
	OtherMale	1.924142	1.734381	0.73	0.468	.3288379	11.25881

**2006-07**

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
> Male LatinoMale OtherMale if validcert2007a==1, rrr iter(20)
```

```
Iteration 0: log likelihood = -25486.465
Iteration 1: log likelihood = -24301.514
Iteration 2: log likelihood = -24279.606
Iteration 3: log likelihood = -24279.568
Iteration 4: log likelihood = -24279.568
```

```
Multinomial logistic regression      Number of obs   =      23874
LR chi2(38)                        =      2413.79
Prob > chi2                         =      0.0000
Pseudo R2                          =      0.0474

Log likelihood = -24279.568
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013292	.0024718	5.41	0.000	1.008459	1.018149
	exper	1.004055	.003052	1.33	0.183	.9980912	1.010055
	male	1.187096	.0487903	4.17	0.000	1.095219	1.286688
	age	.9745761	.0024401	-10.29	0.000	.9698053	.9793703
pyacctrati~2		.9783835	.0249506	-0.86	0.391	.9306832	1.028529
pyperstudlep		1.007447	.0012815	5.83	0.000	1.004938	1.009962
pyperstude~s		.9927348	.001258	-5.75	0.000	.9902723	.9952035
pyperstudw~e		.9919012	.0034031	-2.37	0.018	.9852535	.9985937
pyperstudb~k		.9946336	.0038503	-1.39	0.165	.9871158	1.002209
pyperstudl~o		.9953357	.0036334	-1.28	0.200	.9882398	1.002483
pyperstudm~e		1.00434	.0021271	2.04	0.041	1.00018	1.008518
	Asian	.8380183	.207294	-0.71	0.475	.516058	1.360845
	Black	1.023204	.0620259	0.38	0.705	.90858	1.152289
	Latino	1.407185	.0773964	6.21	0.000	1.263381	1.567357
	Other	.8530675	.161853	-0.84	0.402	.5881451	1.237321
	AsianMale	1.161677	.6756472	1.15	0.250	.7127411	3.667455
	BlackMale	1.766184	.1728201	5.81	0.000	1.457963	2.139565
	LatinoMale	1.14844	.0923465	1.72	0.085	.9809866	1.344479
	OtherMale	1.736233	.5263119	1.82	0.069	.9584725	3.145115
3							
	tenure	1.015979	.0023845	6.75	0.000	1.011317	1.020664
	exper	1.055519	.0035184	16.21	0.000	1.048646	1.062438
	male	1.428321	.0612768	8.31	0.000	1.313132	1.553615
	age	.9787949	.0028724	-7.30	0.000	.9731812	.984441
pyacctrati~2		1.513138	.0409595	15.30	0.000	1.434951	1.595585
pyperstudlep		1.009731	.0013873	7.05	0.000	1.007016	1.012454
pyperstude~s		1.018476	.0013866	13.45	0.000	1.015762	1.021197
pyperstudw~e		1.011815	.0041219	2.88	0.004	1.003768	1.019926
pyperstudb~k		.9981723	.0045729	-0.40	0.690	.9892497	1.007175
pyperstudl~o		.9938696	.0042862	-1.43	0.154	.9855042	1.002306
pyperstudm~e		.9947786	.0025908	-2.01	0.044	.9897136	.9998695
	Asian	.8191581	.2444179	-0.67	0.504	.4564453	1.470099
	Black	.7891653	.0569786	-3.28	0.001	.6850311	.9091293
	Latino	1.200705	.0742749	2.96	0.003	1.063608	1.355474
	Other	.6861283	.158317	-1.63	0.103	.4365152	1.078478
	AsianMale	1.212673	.6145599	0.38	0.704	.4491288	3.274287
	BlackMale	1.566289	.182814	3.84	0.000	1.246009	1.968894
	LatinoMale	1.029866	.0931741	0.33	0.745	.8625234	1.229676
	OtherMale	1.192749	.4557025	0.46	0.645	.5640772	2.522085

```
. mlogtest, all
```

```
**** Likelihood-ratio tests for independent variables (N=23874)
```

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	55.069	2	0.000
exper	295.192	2	0.000
male	70.798	2	0.000
age	126.623	2	0.000
pyacctrati~2	285.232	2	0.000
pyperstudlep	60.666	2	0.000
pyperstude~s	311.752	2	0.000
pyperstudw~e	22.119	2	0.000
pyperstudb~k	1.938	2	0.379
pyperstudl~o	2.714	2	0.257
pyperstudm~e	12.525	2	0.002
Asian	0.747	2	0.688
Black	13.224	2	0.001
Latino	39.017	2	0.000
Other	2.944	2	0.229
AsianMale	1.325	2	0.515
BlackMale	37.294	2	0.000
LatinoMale	3.144	2	0.208
OtherMale	3.340	2	0.188

\*\*\*\* Wald tests for independent variables (N=23874)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	54.979	2	0.000
exper	278.993	2	0.000
male	70.734	2	0.000
age	124.975	2	0.000
pyacctrati~2	283.616	2	0.000
pyperstudlep	60.052	2	0.000
pyperstude~s	304.169	2	0.000
pyperstudw~e	21.716	2	0.000
pyperstudb~k	1.943	2	0.379
pyperstudl~o	2.709	2	0.258
pyperstudm~e	12.350	2	0.002
Asian	0.745	2	0.689
Black	12.957	2	0.002
Latino	39.007	2	0.000
Other	2.836	2	0.242
AsianMale	1.323	2	0.516
BlackMale	36.739	2	0.000
LatinoMale	3.145	2	0.208
OtherMale	3.334	2	0.189

\*\*\*\* Hausman tests of IIA assumption (N=23874)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-336.093	19	---	---
3	0.436	19	1.000	for Ho

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=23874)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-5023.419	-5016.515	13.807	20	0.840	for Ho
3	-5967.521	-5958.447	18.147	20	0.578	for Ho

\*\*\*\* Wald tests for combining alternatives (N=23874)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1524.505	19	0.000
2-	1	482.582	19	0.000
3-	1	1430.445	19	0.000

\*\*\*\* LR tests for combining alternatives (N=23874)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1695.529	19	0.000
2-	1	496.791	19	0.000
3-	1	1586.713	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
> AsianMale BlackMale LatinoMale OtherMale if validcert2007a==1, rrr iter(20)
```

-----> region = 1

note: OtherMale omitted because of collinearity

```
Iteration 0: log likelihood = -1834.7109
Iteration 1: log likelihood = -1743.9774
Iteration 2: log likelihood = -1740.32
Iteration 3: log likelihood = -1739.6273
Iteration 4: log likelihood = -1739.5283
Iteration 5: log likelihood = -1739.5087
Iteration 6: log likelihood = -1739.5042
Iteration 7: log likelihood = -1739.5031
Iteration 8: log likelihood = -1739.5029
Iteration 9: log likelihood = -1739.5028
```

```
Multinomial logistic regression      Number of obs   =      1702
LR chi2(36)                         =      190.42
Prob > chi2                         =      0.0000
Pseudo R2                           =      0.0519
```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1					
tenure	.982265	.0084758	-2.07	0.038	.9657924 .9990186
exper	1.023804	.0130084	1.85	0.064	.9986224 1.04962
male	.760196	.2296459	-0.91	0.364	.4205214 1.374242
age	1.006777	.0109141	0.62	0.533	.985611 1.028397

pyacctrati~2		1.150767	.1087493	1.49	0.137	.9561973	1.384928
pyperstudlep		.9940288	.0029964	-1.99	0.047	.9881733	.9999191
pyperstude~s		.9941995	.0066278	-0.87	0.383	.9812938	1.007275
pyperstudw~e		.9562127	.0681301	-0.63	0.530	.8315846	1.099519
pyperstudb~k		.7237262	.158459	-1.48	0.140	.4711993	1.111588
pyperstudl~o		.9612209	.0626393	-0.61	0.544	.8459671	1.092177
pyperstudm~e		1.015752	.009765	1.63	0.104	.9967921	1.035072
Asian		3.515543	3.95684	1.12	0.264	.3872059	31.91854
Black		1492957	1.29e+09	0.02	0.987	0	.
Latino		.8001954	.1736166	-1.03	0.304	.5230134	1.224276
Other		.475753	.37737	-0.94	0.349	.1005087	2.251954
AsianMale		.3348458	1634.136	-0.00	1.000	0	.
BlackMale		5.72e-07	.0004936	-0.02	0.987	0	.
LatinoMale		.9778158	.3217454	-0.07	0.946	.5130662	1.863548
OtherMale		(omitted)					

2		(base outcome)					
3							
tenure		1.016772	.010115	1.67	0.095	.9971392	1.036792
exper		1.072745	.0181952	4.14	0.000	1.037669	1.109006
male		.651162	.2388896	-1.17	0.242	.3172573	1.336492
age		.9707833	.0144139	-2.00	0.046	.9429397	.9994492
pyacctrati~2		1.754935	.1893631	5.21	0.000	1.42041	2.168246
pyperstudlep		1.010334	.0034667	3.00	0.003	1.003562	1.017151
pyperstude~s		.9836925	.0077179	-2.10	0.036	.9686815	.9989361
pyperstudw~e		1.151735	.1021124	1.59	0.111	.9680221	1.370313
pyperstudb~k		1.172905	.1065414	1.76	0.079	.98162	1.401465
pyperstudl~o		1.150685	.0925734	1.74	0.081	.9828263	1.347213
pyperstudm~e		1.015552	.0118896	1.32	0.187	.992514	1.039124
Asian		1.68e-06	.001511	-0.01	0.988	0	.
Black		1826483	1.58e+09	0.02	0.987	0	.
Latino		.6920836	.1690415	-1.51	0.132	.4287977	1.117029
Other		.3133165	.3597887	-1.01	0.312	.0330011	2.974665
AsianMale		3.63e+13	1.29e+17	0.01	0.993	0	.
BlackMale		4.86e-13	5.43e-10	-0.03	0.980	0	.
LatinoMale		1.50444	.597131	1.03	0.303	.6910706	3.275121
OtherMale		(omitted)					

-----> region = 2

```

Iteration 0: log likelihood = -775.56601
Iteration 1: log likelihood = -723.39046
Iteration 2: log likelihood = -721.61908
Iteration 3: log likelihood = -721.32538
Iteration 4: log likelihood = -721.28022
Iteration 5: log likelihood = -721.2693
Iteration 6: log likelihood = -721.26701
Iteration 7: log likelihood = -721.26653
Iteration 8: log likelihood = -721.26642
Iteration 9: log likelihood = -721.26639

```

Multinomial logistic regression	Number of obs	=	746
	LR chi2(38)	=	108.60
	Prob > chi2	=	0.0000
Log likelihood = -721.26639	Pseudo R2	=	0.0700

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.007901	.0146969	0.54	0.589	.9795032 1.037122
exper		.9812759	.0184092	-1.01	0.314	.9458498 1.018029

male		1.253774	.3593778	0.79	0.430	.7148805	2.198897
age		.969746	.0151249	-1.97	0.049	.9405503	.9998479
pyacctrati~2		1.031093	.1967922	0.16	0.873	.7093157	1.498844
pyperstudlep		.969034	.0175636	-1.74	0.083	.9352142	1.004077
pyperstude~s		1.015894	.0077603	2.06	0.039	1.000797	1.031218
pyperstudw~e		.7730909	.0640084	-3.11	0.002	.6572869	.9092978
pyperstudb~k		.730417	.0695978	-3.30	0.001	.6059882	.880395
pyperstudl~o		.779346	.0624188	-3.11	0.002	.6661262	.9118095
pyperstudm~e		1.009727	.0134734	0.73	0.468	.9836621	1.036483
Asian		5.47e-07	.0007319	-0.01	0.991	0	.
Black		1.578391	1.261256	0.57	0.568	.3296397	7.557704
Latino		1.46639	.3812293	1.47	0.141	.8809594	2.440863
Other		8.92e-07	.0006793	-0.02	0.985	0	.
AsianMale		3.74618	6873.146	0.00	0.999	0	.
BlackMale		1.608311	1.921502	0.40	0.691	.154672	16.72355
LatinoMale		1.157988	.4620294	0.37	0.713	.5297619	2.531207
OtherMale		2744302	2.09e+09	0.02	0.984	0	.
-----							
3							
tenure		1.000521	.0126198	0.04	0.967	.9760902	1.025564
exper		1.037802	.0187518	2.05	0.040	1.001693	1.075214
male		.7947042	.2272434	-0.80	0.422	.4537398	1.391888
age		.9846997	.0157645	-0.96	0.336	.9542815	1.016088
pyacctrati~2		1.691065	.3073453	2.89	0.004	1.184288	2.4147
pyperstudlep		.9994665	.0164035	-0.03	0.974	.9678279	1.032139
pyperstude~s		1.020061	.007817	2.59	0.010	1.004855	1.035498
pyperstudw~e		.9782187	.0831559	-0.26	0.796	.8280899	1.155565
pyperstudb~k		.9024022	.0905663	-1.02	0.306	.7412632	1.09857
pyperstudl~o		.9683291	.0799334	-0.39	0.697	.8236795	1.138381
pyperstudm~e		.9796383	.0155761	-1.29	0.196	.9495805	1.010648
Asian		8.01e-07	.0010917	-0.01	0.992	0	.
Black		2.480014	1.816687	1.24	0.215	.5900958	10.42283
Latino		.9644965	.2450965	-0.14	0.887	.5861308	1.587109
Other		.5928863	.6838155	-0.45	0.650	.0618335	5.684851
AsianMale		5068532	6.91e+09	0.01	0.991	0	.
BlackMale		.7084892	.9976604	-0.24	0.807	.0448448	11.1932
LatinoMale		2.496574	1.002855	2.28	0.023	1.136118	5.486123
OtherMale		2.28e-06	.0019959	-0.01	0.988	0	.
-----							

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -311.41072  
Iteration 1: log likelihood = -274.78566  
Iteration 2: log likelihood = -273.40009  
Iteration 3: log likelihood = -273.22034  
Iteration 4: log likelihood = -273.181  
Iteration 5: log likelihood = -273.17276  
Iteration 6: log likelihood = -273.1714  
Iteration 7: log likelihood = -273.17108  
Iteration 8: log likelihood = -273.17101  
Iteration 9: log likelihood = -273.17099

Multinomial logistic regression	Number of obs	=	286
	LR chi2(32)	=	76.48
	Prob > chi2	=	0.0000
Log likelihood = -273.17099	Pseudo R2	=	0.1228

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				

```

2
  tenure | 1.075995 .0277491 2.84 0.005 1.02296 1.13178
    exper | .8701716 .0292194 -4.14 0.000 .8147465 .9293672
      male | 2.35351 .9535824 2.11 0.035 1.063728 5.207164
      age | 1.047153 .0285062 1.69 0.091 .992746 1.104541
pyacctrati~2 | 1.523267 .4342562 1.48 0.140 .8711917 2.663412
pyperstudlep | 1.009832 .0443537 0.22 0.824 .9265373 1.100615
pyperstude~s | .9932241 .0213822 -0.32 0.752 .9521876 1.036029
pyperstudw~e | .8887728 .0615507 -1.70 0.089 .7759648 1.01798
pyperstudb~k | .9402442 .0665154 -0.87 0.384 .8185107 1.080083
pyperstudl~o | .9149871 .0653678 -1.24 0.214 .7954339 1.052509
pyperstudm~e | .9988243 .0258099 -0.05 0.964 .9494974 1.050714
    Asian | (omitted)
    Black | 1.662395 1.473939 0.57 0.566 .2924344 9.450179
    Latino | 2.378522 1.662598 1.24 0.215 .6043844 9.360544
    Other | 552965 4.63e+08 0.02 0.987 0 .
    AsianMale | (omitted)
    BlackMale | 1767073 1.19e+09 0.02 0.983 0 .
    LatinoMale | 1.141324 1.223034 0.12 0.902 .1397228 9.322883
    OtherMale | (omitted)

```

```

3
  tenure | 1.024135 .0213204 1.15 0.252 .9831892 1.066787
    exper | .9717762 .0303312 -0.92 0.359 .91411 1.03308
      male | 1.320171 .4789777 0.77 0.444 .6483385 2.688183
      age | 1.000475 .0277645 0.02 0.986 .9475112 1.0564
pyacctrati~2 | 1.852049 .4953668 2.30 0.021 1.096433 3.128405
pyperstudlep | 1.042324 .0392194 1.10 0.271 .9682213 1.122098
pyperstude~s | 1.027508 .0197263 1.41 0.158 .9895637 1.066908
pyperstudw~e | .9921593 .0641816 -0.12 0.903 .8740135 1.126275
pyperstudb~k | .9834503 .0652874 -0.25 0.802 .8634645 1.120109
pyperstudl~o | .9754841 .064964 -0.37 0.709 .8561167 1.111495
pyperstudm~e | .9873129 .0267965 -0.47 0.638 .9361652 1.041255
    Asian | (omitted)
    Black | 1.829552 1.422874 0.78 0.437 .3984289 8.401151
    Latino | 1.911166 1.305406 0.95 0.343 .5010633 7.289613
    Other | 2736232 2.29e+09 0.02 0.986 0 .
    AsianMale | (omitted)
    BlackMale | 1043530 7.00e+08 0.02 0.984 0 .
    LatinoMale | 1.559664 1.616735 0.43 0.668 .2044949 11.89542
    OtherMale | (omitted)

```

-----> region = 4

```

Iteration 0: log likelihood = -5276.2092
Iteration 1: log likelihood = -4970.2098
Iteration 2: log likelihood = -4956.6735
Iteration 3: log likelihood = -4956.6122
Iteration 4: log likelihood = -4956.6122

```

```

Multinomial logistic regression      Number of obs   =      5116
                                     LR chi2(38)        =      639.19
                                     Prob > chi2         =      0.0000
Log likelihood = -4956.6122          Pseudo R2        =      0.0606

```

```

-----+-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          | (base outcome)
-----+-----
2
  tenure | 1.028081 .0055185 5.16 0.000 1.017322 1.038954
    exper | 1.024862 .007066 3.56 0.000 1.011106 1.038806
      male | 1.099744 .1004153 1.04 0.298 .9195386 1.315264
      age | .9651346 .0053766 -6.37 0.000 .954654 .9757302

```



pyacctrati~2		.9564739	.0524259	-0.81	0.417	.8590479	1.064949
pyperstudlep		1.001409	.0029634	0.48	0.634	.9956182	1.007235
pyperstude~s		.9949131	.0036139	-1.40	0.160	.9878552	1.002022
pyperstudw~e		1.003085	.0053094	0.58	0.561	.9927325	1.013545
pyperstudb~k		.9999815	.0066359	-0.00	0.998	.9870595	1.013073
pyperstudl~o		1.005465	.0064238	0.85	0.394	.9929528	1.018134
pyperstudm~e		1.008065	.0051648	1.57	0.117	.9979929	1.018239
Asian		.925537	.3564792	-0.20	0.841	.4350564	1.968983
Black		.8552201	.0881304	-1.52	0.129	.698814	1.046633
Latino		1.44233	.1830584	2.89	0.004	1.124686	1.849686
Other		.9091486	.3329661	-0.26	0.795	.4434976	1.863711
AsianMale		1.153461	.7582296	0.22	0.828	.3180269	4.183519
BlackMale		2.070095	.3431144	4.39	0.000	1.495909	2.864675
LatinoMale		1.062637	.2252295	0.29	0.774	.7014066	1.609904
OtherMale		1.250357	.9622407	0.29	0.772	.2766767	5.650613
-----							
3							
tenure		1.043152	.0062865	7.01	0.000	1.030904	1.055547
exper		1.072496	.009481	7.92	0.000	1.054074	1.091241
male		1.216704	.1359557	1.76	0.079	.9773967	1.514604
age		.965663	.007315	-4.61	0.000	.9514318	.980107
pyacctrati~2		1.443816	.0949253	5.59	0.000	1.269254	1.642385
pyperstudlep		1.013143	.0036771	3.60	0.000	1.005961	1.020375
pyperstude~s		1.026952	.0047193	5.79	0.000	1.017744	1.036244
pyperstudw~e		1.016358	.0070492	2.34	0.019	1.002635	1.030269
pyperstudb~k		.997246	.0087722	-0.31	0.754	.9802001	1.014588
pyperstudl~o		.9857205	.0082034	-1.73	0.084	.9697726	1.001931
pyperstudm~e		.9982574	.007057	-0.25	0.805	.9845213	1.012185
Asian		.8657756	.4351653	-0.29	0.774	.32327	2.318704
Black		.4873916	.0682195	-5.13	0.000	.370456	.6412384
Latino		1.219612	.1902007	1.27	0.203	.8984112	1.655648
Other		.6932981	.3597785	-0.71	0.480	.2507261	1.917081
AsianMale		.3230491	.3874461	-0.94	0.346	.0307884	3.389609
BlackMale		2.081371	.462687	3.30	0.001	1.346257	3.217887
LatinoMale		1.244262	.3305872	0.82	0.411	.7391908	2.094435
OtherMale		1.506639	1.839461	0.34	0.737	.1376493	16.4909
-----							

-----> region = 5

note: OtherMale omitted because of collinearity

```

Iteration 0: log likelihood = -492.80108
Iteration 1: log likelihood = -447.26245
Iteration 2: log likelihood = -445.36998
Iteration 3: log likelihood = -445.03678
Iteration 4: log likelihood = -444.98215
Iteration 5: log likelihood = -444.96963
Iteration 6: log likelihood = -444.9666
Iteration 7: log likelihood = -444.966
Iteration 8: log likelihood = -444.9659
Iteration 9: log likelihood = -444.96588

```

Multinomial logistic regression	Number of obs	=	482
	LR chi2(36)	=	95.67
	Prob > chi2	=	0.0000
Log likelihood = -444.96588	Pseudo R2	=	0.0971

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----					
1	(base outcome)				
-----					
2					
tenure	1.02167	.0187368	1.17	0.242	.9855983 1.059061
exper	1.007563	.0264937	0.29	0.774	.9569518 1.060851
male	2.655277	.7586414	3.42	0.001	1.516745 4.648441

age		.9525199	.0209789	-2.21	0.027	.9122768	.9945382
pyaccrati~2		.9792178	.2369929	-0.09	0.931	.6093534	1.573582
pyperstudlep		.9427666	.0366795	-1.51	0.130	.8735487	1.017469
pyperstude~s		1.003198	.0101441	0.32	0.752	.9835119	1.023279
pyperstudw~e		1.121698	.0700918	1.84	0.066	.9924001	1.267842
pyperstudb~k		1.11663	.0711959	1.73	0.084	.9854556	1.265265
pyperstudl~o		1.183842	.0988015	2.02	0.043	1.005203	1.394227
pyperstudm~e		1.00091	.0284549	0.03	0.974	.9466646	1.058263
Asian		4.94e-07	.0008537	-0.01	0.993	0	.
Black		.9643787	.4323511	-0.08	0.936	.400531	2.321983
Latino		1.572657	1.914534	0.37	0.710	.1446745	17.09529
Other		1.707621	2.211908	0.41	0.680	.1348395	21.62549
AsianMale		1350763	2.33e+09	0.01	0.993	0	.
BlackMale		1.149623	.7124881	0.22	0.822	.3412056	3.873419
LatinoMale		1732810	1.56e+09	0.02	0.987	0	.
OtherMale		(omitted)					
-----							
3							
tenure		1.013618	.0159535	0.86	0.390	.9828268	1.045373
exper		1.078907	.0290688	2.82	0.005	1.023411	1.137412
male		2.411241	.6932396	3.06	0.002	1.372516	4.236077
age		.9510426	.022301	-2.14	0.032	.9083226	.9957717
pyaccrati~2		1.835878	.3917172	2.85	0.004	1.208439	2.789092
pyperstudlep		.9580626	.0364865	-1.12	0.261	.8891541	1.032311
pyperstude~s		1.019642	.0099606	1.99	0.046	1.000306	1.039353
pyperstudw~e		1.065078	.0552962	1.21	0.225	.9620313	1.179163
pyperstudb~k		1.046946	.05563	0.86	0.388	.9433987	1.161859
pyperstudl~o		1.084951	.0765269	1.16	0.248	.9448672	1.245803
pyperstudm~e		.9536106	.0289365	-1.57	0.117	.8985496	1.012046
Asian		7.19e-07	.0012318	-0.01	0.993	0	.
Black		2.759766	1.17452	2.39	0.017	1.198429	6.355246
Latino		2.17e-06	.0020512	-0.01	0.989	0	.
Other		2.52e-06	.0029342	-0.01	0.991	0	.
AsianMale		.7690078	1767.434	-0.00	1.000	0	.
BlackMale		.832984	.5095151	-0.30	0.765	.2511761	2.762453
LatinoMale		1.85e+12	2.41e+15	0.02	0.983	0	.
OtherMale		(omitted)					
-----							

-----> region = 6

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -882.52151
Iteration 1: log likelihood = -831.39381
Iteration 2: log likelihood = -829.83189
Iteration 3: log likelihood = -829.69892
Iteration 4: log likelihood = -829.67207
Iteration 5: log likelihood = -829.66644
Iteration 6: log likelihood = -829.66517
Iteration 7: log likelihood = -829.66486
Iteration 8: log likelihood = -829.6648
Iteration 9: log likelihood = -829.66479

```

Multinomial logistic regression	Number of obs	=	826
	LR chi2(36)	=	105.71
	Prob > chi2	=	0.0000
Log likelihood = -829.66479	Pseudo R2	=	0.0599

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9890682	.01392	-0.78	0.435	.9621584 1.016731
exper		1.006654	.015784	0.42	0.672	.976189 1.038071

male		1.179576	.2290286	0.85	0.395	.8062225	1.725825
age		.975818	.0127253	-1.88	0.060	.951193	1.001081
pyacctrati~2		1.139633	.1516657	0.98	0.326	.8779798	1.479263
pyperstudlep		1.021153	.0156254	1.37	0.171	.9909824	1.052242
pyperstude~s		.9990644	.0070899	-0.13	0.895	.9852646	1.013058
pyperstudw~e		1.042222	.0336924	1.28	0.201	.9782349	1.110395
pyperstudb~k		1.038471	.0362579	1.08	0.280	.9697835	1.112023
pyperstudl~o		1.031801	.0340054	0.95	0.342	.9672583	1.10065
pyperstudm~e		.9940101	.0125537	-0.48	0.634	.9697074	1.018922
Asian		2.211764	3.23502	0.54	0.587	.1258157	38.88145
Black		1.16013	.5036391	0.34	0.732	.4954277	2.716644
Latino		.487756	.3380304	-1.04	0.300	.1253987	1.897197
Other		5.18e-07	.0006288	-0.01	0.990	0	.
AsianMale		(omitted)					
BlackMale		1.909269	1.349984	0.91	0.360	.4775399	7.633519
LatinoMale		.9099858	.8333559	-0.10	0.918	.1511864	5.477173
OtherMale		2670346	3.24e+09	0.01	0.990	0	.
-----							
3							
tenure		.9979169	.0129807	-0.16	0.873	.9727968	1.023686
exper		1.077478	.0195105	4.12	0.000	1.039908	1.116404
male		2.2375	.4530674	3.98	0.000	1.504544	3.327525
age		.9563225	.0153282	-2.79	0.005	.9267467	.9868422
pyacctrati~2		1.667318	.2320524	3.67	0.000	1.26926	2.190212
pyperstudlep		1.018603	.0163469	1.15	0.251	.9870622	1.051151
pyperstude~s		1.027894	.0077043	3.67	0.000	1.012904	1.043105
pyperstudw~e		1.017065	.0299108	0.58	0.565	.960098	1.077411
pyperstudb~k		1.017492	.0316966	0.56	0.578	.9572259	1.081552
pyperstudl~o		.989042	.0294897	-0.37	0.712	.9328996	1.048563
pyperstudm~e		.9872202	.0168885	-0.75	0.452	.9546681	1.020882
Asian		4.67e-06	.0044635	-0.01	0.990	0	.
Black		1.20472	.5626899	0.40	0.690	.4822972	3.009247
Latino		1.152058	.7457792	0.22	0.827	.3239318	4.097279
Other		5.94e-07	.000716	-0.01	0.991	0	.
AsianMale		(omitted)					
BlackMale		1.074731	.8228141	0.09	0.925	.2396735	4.819252
LatinoMale		.3182958	.2984379	-1.22	0.222	.0506681	1.999528
OtherMale		.8840361	1370.722	-0.00	1.000	0	.
-----							

-----> region = 7

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1103.7188  
Iteration 1: log likelihood = -1037.5323  
Iteration 2: log likelihood = -1035.6844  
Iteration 3: log likelihood = -1035.6081  
Iteration 4: log likelihood = -1035.5964  
Iteration 5: log likelihood = -1035.5951  
Iteration 6: log likelihood = -1035.5949  
Iteration 7: log likelihood = -1035.5948  
Iteration 8: log likelihood = -1035.5948

Multinomial logistic regression	Number of obs	=	1079
	LR chi2(34)	=	136.25
	Prob > chi2	=	0.0000
Log likelihood = -1035.5948	Pseudo R2	=	0.0617

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9802481	.0121509	-1.61	0.108	.9567199 1.004355

exper		1.004118	.0149425	0.28	0.782	.9752545	1.033836
male		1.076635	.2032918	0.39	0.696	.7436066	1.558811
age		.9829062	.012481	-1.36	0.175	.9587459	1.007675
pyaccrati~2		.8410161	.1386238	-1.05	0.294	.6088378	1.161735
pyperstudlep		.9796163	.0145354	-1.39	0.165	.9515377	1.008523
pyperstude~s		.9912501	.0081114	-1.07	0.283	.9754789	1.007276
pyperstudw~e		.9035384	.0851775	-1.08	0.282	.751109	1.086902
pyperstudb~k		.9045985	.0866512	-1.05	0.295	.7497553	1.091421
pyperstudl~o		.9271687	.088736	-0.79	0.429	.7685874	1.11847
pyperstudm~e		.9906734	.0128232	-0.72	0.469	.9658565	1.016128
Asian		(omitted)					
Black		1.200705	.4163713	0.53	0.598	.6085033	2.369242
Latino		.5840457	.6497638	-0.48	0.629	.0659882	5.169248
Other		.3962869	.4335717	-0.85	0.398	.0464219	3.38296
AsianMale		(omitted)					
BlackMale		2.377941	1.166451	1.77	0.077	.9092039	6.219289
LatinoMale		2.73543	3.50778	0.78	0.433	.2215578	33.77258
OtherMale		5.64e-06	.0047426	-0.01	0.989	0	.
-----							
3							
tenure		.9880647	.0099095	-1.20	0.231	.9688321	1.007679
exper		1.053151	.0145117	3.76	0.000	1.02509	1.081981
male		1.582954	.2696744	2.70	0.007	1.133591	2.210447
age		.9819259	.0122186	-1.47	0.143	.9582677	1.006168
pyaccrati~2		1.606992	.2281898	3.34	0.001	1.216591	2.122671
pyperstudlep		1.025571	.0149571	1.73	0.083	.996671	1.05531
pyperstude~s		1.028813	.007474	3.91	0.000	1.014268	1.043567
pyperstudw~e		.8742277	.0780309	-1.51	0.132	.7339204	1.041358
pyperstudb~k		.8632521	.0782455	-1.62	0.105	.7227439	1.031076
pyperstudl~o		.8494444	.0772398	-1.79	0.073	.7107803	1.01516
pyperstudm~e		.9836457	.0120057	-1.35	0.177	.9603942	1.00746
Asian		(omitted)					
Black		1.208483	.390519	0.59	0.558	.6414699	2.276694
Latino		1.238323	.9330768	0.28	0.777	.282783	5.422687
Other		.4364365	.4821012	-0.75	0.453	.0500792	3.80351
AsianMale		(omitted)					
BlackMale		1.660824	.7792487	1.08	0.280	.6621327	4.165837
LatinoMale		.7007556	.7132973	-0.35	0.727	.0953083	5.152315
OtherMale		3.15e-06	.0021755	-0.02	0.985	0	.
-----							

-----> region = 8

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -390.52114  
Iteration 1: log likelihood = -346.69628  
Iteration 2: log likelihood = -344.63841  
Iteration 3: log likelihood = -344.4943  
Iteration 4: log likelihood = -344.46174  
Iteration 5: log likelihood = -344.45418  
Iteration 6: log likelihood = -344.45261  
Iteration 7: log likelihood = -344.45234  
Iteration 8: log likelihood = -344.45231

Multinomial logistic regression	Number of obs	=	374
	LR chi2(34)	=	92.14
	Prob > chi2	=	0.0000
Log likelihood = -344.45231	Pseudo R2	=	0.1180

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					

tenure		1.026785	.0243902	1.11	0.266	.9800773	1.07572
exper		.9895176	.0295855	-0.35	0.725	.9331974	1.049237
male		3.180172	1.099246	3.35	0.001	1.615204	6.261431
age		.9674479	.0259005	-1.24	0.216	.9179928	1.019567
pyacctrati~2		1.331335	.3890777	0.98	0.327	.7508042	2.360738
pyperstudlep		1.016138	.0343343	0.47	0.636	.9510241	1.085711
pyperstude~s		.9772561	.0135071	-1.66	0.096	.9511379	1.004091
pyperstudw~e		1.054815	.1571607	0.36	0.720	.7876856	1.412537
pyperstudb~k		1.111253	.1675534	0.70	0.484	.8269324	1.493332
pyperstudl~o		1.071401	.1638356	0.45	0.652	.7939414	1.445823
pyperstudm~e		.9651693	.0341944	-1.00	0.317	.9004235	1.034571
Asian		1.685138	4532.463	0.00	1.000	0	.
Black		3.493981	1.89391	2.31	0.021	1.207612	10.10913
Latino		3157121	1.69e+09	0.03	0.978	0	.
Other		6.21076	8.139393	1.39	0.163	.476013	81.03463
AsianMale		(omitted)					
BlackMale		1.077823	1.373665	0.06	0.953	.0886557	13.10352
LatinoMale		(omitted)					
OtherMale		1.06e-07	.0001703	-0.01	0.992	0	.
-----							
3							
tenure		.9949227	.0182491	-0.28	0.781	.9597904	1.031341
exper		.9983678	.0227151	-0.07	0.943	.9548251	1.043896
male		2.634385	.754642	3.38	0.001	1.502607	4.61863
age		1.020593	.0222916	0.93	0.351	.9778238	1.065232
pyacctrati~2		2.245745	.5627183	3.23	0.001	1.374277	3.669837
pyperstudlep		1.007357	.032858	0.22	0.822	.9449715	1.07386
pyperstude~s		1.034942	.0117525	3.02	0.002	1.012162	1.058235
pyperstudw~e		1.056854	.1189329	0.49	0.623	.8476671	1.317665
pyperstudb~k		1.045447	.1177807	0.39	0.693	.838311	1.303764
pyperstudl~o		1.029416	.1177876	0.25	0.800	.8226114	1.28821
pyperstudm~e		1.015281	.0216736	0.71	0.477	.9736783	1.058662
Asian		717158.7	1.06e+09	0.01	0.993	0	.
Black		1.873434	1.061556	1.11	0.268	.6170444	5.68801
Latino		1146072	6.15e+08	0.03	0.979	0	.
Other		6.010256	7.623059	1.41	0.157	.5003553	72.19506
AsianMale		(omitted)					
BlackMale		1.842559	2.434457	0.46	0.644	.1382872	24.55051
LatinoMale		(omitted)					
OtherMale		1.14e-07	.0001463	-0.01	0.990	0	.
-----							

-----> region = 9

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -212.83237  
Iteration 1: log likelihood = -191.3347  
Iteration 2: log likelihood = -190.48847  
Iteration 3: log likelihood = -189.64615  
Iteration 4: log likelihood = -189.58673  
Iteration 5: log likelihood = -189.5764  
Iteration 6: log likelihood = -189.57407  
Iteration 7: log likelihood = -189.5735  
Iteration 8: log likelihood = -189.57339  
Iteration 9: log likelihood = -189.57337  
Iteration 10: log likelihood = -189.57337

Multinomial logistic regression	Number of obs	=	201
	LR chi2(34)	=	46.52
	Prob > chi2	=	0.0746
Log likelihood = -189.57337	Pseudo R2	=	0.1093

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						

```

1      |
      tenure | 1.008883 .0247019 0.36 0.718 .9616113 1.058478
      exper | .981792 .0356091 -0.51 0.612 .9144223 1.054125
      male | .8405395 .3125603 -0.47 0.640 .4055394 1.74214
      age | .9826863 .0315775 -0.54 0.587 .9227043 1.046568
pyacctrati~2 | .7994681 .2425626 -0.74 0.461 .4411054 1.448972
pyperstudlep | .9817811 .0672661 -0.27 0.788 .8584108 1.122882
pyperstude~s | .9809746 .0152605 -1.23 0.217 .951516 1.011345
pyperstudw~e | .8239055 .0949759 -1.68 0.093 .6572869 1.032761
pyperstudb~k | .8228838 .0990439 -1.62 0.105 .6499594 1.041816
pyperstudl~o | .8507765 .1016271 -1.35 0.176 .67319 1.07521
pyperstudm~e | .9923411 .0337131 -0.23 0.821 .9284164 1.060667
      Asian | 1.223755 22496.48 0.00 1.000 0 .
      Black | 1.20e-07 .0003021 -0.01 0.995 0 .
      Latino | 4.02e-08 .0002188 -0.00 0.998 0 .
      Other | .8001513 14709.31 -0.00 1.000 0 .
      AsianMale | (omitted)
      BlackMale | 1.77e+14 9.05e+17 0.01 0.995 0 .
      LatinoMale | (omitted)
      OtherMale | 9.01e-08 .0016856 -0.00 0.999 0 .
-----

```

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2      |
      tenure | .9749348 .030447 -0.81 0.416 .9170494 1.036474
      exper | .9948941 .0434015 -0.12 0.907 .9133638 1.083702
      male | .4719848 .2168488 -1.63 0.102 .1918014 1.16146
      age | .9875146 .0365265 -0.34 0.734 .9184574 1.061764
pyacctrati~2 | .7991891 .3131284 -0.57 0.567 .3708022 1.72249
pyperstudlep | .9605155 .0910027 -0.43 0.671 .7977347 1.156512
pyperstude~s | .9649929 .0197605 -1.74 0.082 .9270298 1.004511
pyperstudw~e | .7422941 .101738 -2.17 0.030 .5674287 .971048
pyperstudb~k | .7775124 .1097132 -1.78 0.075 .5896523 1.025224
pyperstudl~o | .779444 .1117941 -1.74 0.082 .5884353 1.032455
pyperstudm~e | 1.004716 .0432245 0.11 0.913 .9234714 1.093109
      Asian | 8.41e+08 1.03e+13 0.00 0.999 0 .
      Black | .3819056 .5454507 -0.67 0.500 .0232402 6.275831
      Latino | 1.15e-07 .0007967 -0.00 0.998 0 .
      Other | 4.39e+08 5.37e+12 0.00 0.999 0 .
      AsianMale | (omitted)
      BlackMale | 2.46e+08 1.09e+12 0.00 0.997 0 .
      LatinoMale | (omitted)
      OtherMale | 3.12e-09 .0000382 -0.00 0.999 0 .
-----

```

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3      | (base outcome)
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-----> region = 10

```

```

Iteration 0: log likelihood = -3788.3914
Iteration 1: log likelihood = -3528.4081
Iteration 2: log likelihood = -3520.4742
Iteration 3: log likelihood = -3520.4481
Iteration 4: log likelihood = -3520.4481

```

```

Multinomial logistic regression      Number of obs   =      3548
                                      LR chi2(38)        =      535.89
                                      Prob > chi2         =      0.0000
Log likelihood = -3520.4481          Pseudo R2        =      0.0707

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1      | (base outcome)
-----+-----
2      |
      tenure | 1.021348      .0066877      3.23  0.001      1.008324      1.03454
      exper | 1.019714      .0084023      2.37  0.018      1.003378      1.036316

```

male		1.465644	.156257	3.59	0.000	1.189266	1.806249
age		.9595289	.0064344	-6.16	0.000	.9470002	.9722233
pyacctrati~2		1.004303	.0652189	0.07	0.947	.8842765	1.140621
pyperstudlep		1.013266	.0047358	2.82	0.005	1.004026	1.022591
pyperstude~s		.9885552	.0036878	-3.09	0.002	.9813536	.9958096
pyperstudw~e		.9917196	.0079215	-1.04	0.298	.9763147	1.007368
pyperstudb~k		.9972153	.0087135	-0.32	0.750	.9802826	1.014441
pyperstudl~o		.9945233	.0084785	-0.64	0.519	.9780438	1.01128
pyperstudm~e		1.004242	.0068771	0.62	0.536	.9908536	1.017812
Asian		1.474652	.872192	0.66	0.511	.4626377	4.700434
Black		1.022031	.1355655	0.16	0.870	.7880573	1.325471
Latino		1.574717	.2918785	2.45	0.014	1.095039	2.264515
Other		1.175326	.5214454	0.36	0.716	.492622	2.804159
AsianMale		1.376997	1.316069	0.33	0.738	.2115413	8.963365
BlackMale		1.387878	.2967644	1.53	0.125	.91273	2.110379
LatinoMale		1.359046	.3900886	1.07	0.285	.7743064	2.385369
OtherMale		1.055275	.6876383	0.08	0.934	.294243	3.784649
-----							
3							
tenure		1.023755	.0065433	3.67	0.000	1.01101	1.03666
exper		1.091633	.0106703	8.97	0.000	1.070919	1.112748
male		1.404794	.167592	2.85	0.004	1.111896	1.774847
age		.955016	.0082278	-5.34	0.000	.9390251	.9712791
pyacctrati~2		1.592822	.1134752	6.53	0.000	1.385244	1.831505
pyperstudlep		1.02381	.0055667	4.33	0.000	1.012957	1.034779
pyperstude~s		1.01353	.0040475	3.37	0.001	1.005628	1.021494
pyperstudw~e		1.00336	.0087121	0.39	0.699	.9864295	1.020582
pyperstudb~k		.9925291	.0096195	-0.77	0.439	.9738532	1.011563
pyperstudl~o		.9764093	.0090354	-2.58	0.010	.95886	.9942799
pyperstudm~e		.9965292	.0081337	-0.43	0.670	.9807143	1.012599
Asian		1.702163	1.158238	0.78	0.434	.4485392	6.459542
Black		.8842023	.1446279	-0.75	0.452	.6416847	1.218377
Latino		2.162152	.4388458	3.80	0.000	1.452514	3.21849
Other		.7642855	.4244902	-0.48	0.628	.2573307	2.269967
AsianMale		.8513866	1.008741	-0.14	0.892	.0834843	8.682575
BlackMale		2.247499	.5652906	3.22	0.001	1.372794	3.679541
LatinoMale		1.616329	.5121569	1.52	0.130	.8685891	3.007773
OtherMale		.9773765	.8323886	-0.03	0.979	.1841327	5.187915
-----							

-----> region = 11

Iteration 0: log likelihood = -2784.7995  
Iteration 1: log likelihood = -2633.0779  
Iteration 2: log likelihood = -2629.4708  
Iteration 3: log likelihood = -2629.3502  
Iteration 4: log likelihood = -2629.3212  
Iteration 5: log likelihood = -2629.3157  
Iteration 6: log likelihood = -2629.3149  
Iteration 7: log likelihood = -2629.3146  
Iteration 8: log likelihood = -2629.3146  
Iteration 9: log likelihood = -2629.3146

Multinomial logistic regression	Number of obs	=	2586
	LR chi2(38)	=	310.97
	Prob > chi2	=	0.0000
Log likelihood = -2629.3146	Pseudo R2	=	0.0558

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.0034	.0078134	0.44	0.663	.9882025 1.018832
exper		1.009347	.008763	1.07	0.284	.9923166 1.026669

male		1.357486	.1498216	2.77	0.006	1.09343	1.685311
age		.9743611	.006808	-3.72	0.000	.9611086	.9877963
pyacctrati~2		1.061091	.0839817	0.75	0.454	.9086213	1.239146
pyperstudlep		.9969099	.0059316	-0.52	0.603	.9853518	1.008604
pyperstude~s		1.001956	.0039074	0.50	0.616	.9943266	1.009643
pyperstudw~e		.9918522	.0111364	-0.73	0.466	.9702637	1.013921
pyperstudb~k		.9927829	.0124291	-0.58	0.563	.9687189	1.017445
pyperstudl~o		.9937859	.011862	-0.52	0.602	.9708066	1.017309
pyperstudm~e		.9978941	.0074434	-0.28	0.777	.9834113	1.01259
Asian		.5801983	.4898994	-0.64	0.519	.1108795	3.035999
Black		1.446105	.2757075	1.93	0.053	.9952081	2.101289
Latino		1.763416	.4203776	2.38	0.017	1.105194	2.813658
Other		.2925936	.1869601	-1.92	0.054	.0836312	1.023674
AsianMale		1301652	8.38e+08	0.02	0.983	0	.
BlackMale		1.005584	.3199507	0.02	0.986	.5389996	1.876068
LatinoMale		1.037744	.3716534	0.10	0.918	.514333	2.093804
OtherMale		2.930993	2.559206	1.23	0.218	.5294014	16.22723

3							
tenure		1.003782	.0075515	0.50	0.616	.9890901	1.018692
exper		1.063605	.0100707	6.51	0.000	1.044049	1.083528
male		1.520871	.1875357	3.40	0.001	1.194352	1.936655
age		.9930973	.0080661	-0.85	0.394	.9774132	1.009033
pyacctrati~2		1.599633	.1383967	5.43	0.000	1.350132	1.895241
pyperstudlep		1.023564	.0070375	3.39	0.001	1.009863	1.03745
pyperstude~s		1.020672	.0045963	4.54	0.000	1.011703	1.02972
pyperstudw~e		1.008239	.0130564	0.63	0.526	.9829712	1.034157
pyperstudb~k		.9970014	.0146554	-0.20	0.838	.9686872	1.026143
pyperstudl~o		.9827067	.0136968	-1.25	0.211	.9562248	1.009922
pyperstudm~e		.9901727	.0093098	-1.05	0.294	.9720928	1.008589
Asian		.8294859	.7199488	-0.22	0.829	.1513604	4.545753
Black		1.413623	.3224849	1.52	0.129	.9039661	2.210626
Latino		1.792453	.4763562	2.20	0.028	1.06472	3.01759
Other		.5176695	.3076447	-1.11	0.268	.1615079	1.659248
AsianMale		1425770	9.18e+08	0.02	0.982	0	.
BlackMale		1.088877	.4259108	0.22	0.828	.5058645	2.343817
LatinoMale		.8416599	.3490415	-0.42	0.678	.3733689	1.897296
OtherMale		1.023973	.9668766	0.03	0.980	.1608978	6.516692

-----> region = 12

```

Iteration 0: log likelihood = -909.92092
Iteration 1: log likelihood = -847.24344
Iteration 2: log likelihood = -845.60814
Iteration 3: log likelihood = -845.36476
Iteration 4: log likelihood = -845.31112
Iteration 5: log likelihood = -845.29984
Iteration 6: log likelihood = -845.29756
Iteration 7: log likelihood = -845.29703
Iteration 8: log likelihood = -845.29691
Iteration 9: log likelihood = -845.29688

```

Multinomial logistic regression	Number of obs	=	839
	LR chi2(38)	=	129.25
	Prob > chi2	=	0.0000
Log likelihood = -845.29688	Pseudo R2	=	0.0710

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9923625	.0122426	-0.62	0.534	.9686551 1.01665
exper		1.040852	.0178145	2.34	0.019	1.006516 1.07636



male		.8036054	.1661821	-1.06	0.290	.5358189	1.205224
age		.9657011	.0143782	-2.34	0.019	.9379275	.9942971
pyacctrati~2		1.258264	.1921926	1.50	0.133	.9327298	1.697415
pyperstudlep		.9913336	.0173678	-0.50	0.619	.9578712	1.025965
pyperstude~s		.9873406	.0082269	-1.53	0.126	.971347	1.003597
pyperstudw~e		.935273	.0529535	-1.18	0.237	.8370374	1.045038
pyperstudb~k		.9536456	.0573064	-0.79	0.430	.8476892	1.072846
pyperstudl~o		.9648227	.0546734	-0.63	0.527	.8634011	1.078158
pyperstudm~e		1.025138	.0131895	1.93	0.054	.99961	1.051317
Asian		9.36e-07	.000713	-0.02	0.985	0	.
Black		.8389588	.2862097	-0.51	0.607	.429887	1.637295
Latino		.7539297	.3931714	-0.54	0.588	.2712893	2.095217
Other		.2130537	.2419033	-1.36	0.173	.0230158	1.972204
AsianMale		3639210	2.77e+09	0.02	0.984	0	.
BlackMale		3.521706	2.106523	2.10	0.035	1.090444	11.37373
LatinoMale		2.496533	1.964824	1.16	0.245	.5338557	11.67484
OtherMale		2.48e+07	1.99e+10	0.02	0.983	0	.
-----							
3							
tenure		.9955513	.0114069	-0.39	0.697	.9734433	1.018161
exper		1.026085	.0161147	1.64	0.101	.9949816	1.05816
male		1.282042	.2549959	1.25	0.212	.8681599	1.893236
age		1.006523	.0140372	0.47	0.641	.979383	1.034415
pyacctrati~2		1.819183	.262853	4.14	0.000	1.370524	2.414716
pyperstudlep		1.021264	.0179568	1.20	0.231	.9866688	1.057072
pyperstude~s		1.01143	.0080541	1.43	0.154	.9957666	1.027339
pyperstudw~e		1.076433	.0630844	1.26	0.209	.9596266	1.207457
pyperstudb~k		1.068466	.0661571	1.07	0.285	.9463597	1.206328
pyperstudl~o		1.061835	.0624238	1.02	0.307	.9462719	1.19151
pyperstudm~e		1.005371	.013767	0.39	0.696	.9787476	1.03272
Asian		.851419	1.089466	-0.13	0.900	.0693356	10.45516
Black		1.342998	.5000435	0.79	0.428	.6473576	2.786162
Latino		.5940224	.415738	-0.74	0.457	.1506861	2.341707
Other		.8724124	.76571	-0.16	0.876	.1561811	4.873211
AsianMale		2.17e-06	.0018944	-0.01	0.988	0	.
BlackMale		2.128935	1.321471	1.22	0.223	.6306745	7.186538
LatinoMale		4.067108	3.828111	1.49	0.136	.6428411	25.73166
OtherMale		1190093	9.56e+08	0.02	0.986	0	.
-----							

-----> region = 13

Iteration 0: log likelihood = -1634.1713  
Iteration 1: log likelihood = -1530.2007  
Iteration 2: log likelihood = -1528.5208  
Iteration 3: log likelihood = -1528.4722  
Iteration 4: log likelihood = -1528.4646  
Iteration 5: log likelihood = -1528.4628  
Iteration 6: log likelihood = -1528.4624  
Iteration 7: log likelihood = -1528.4623  
Iteration 8: log likelihood = -1528.4623

Multinomial logistic regression	Number of obs	=	1499
	LR chi2(38)	=	211.42
	Prob > chi2	=	0.0000
Log likelihood = -1528.4623	Pseudo R2	=	0.0647

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.008685	.0106945	0.82	0.415	.9879402 1.029865
exper		.9774832	.0107964	-2.06	0.039	.95655 1.9988745
male		1.106348	.1750634	0.64	0.523	.8113373 1.508627

age		.9698202	.0091346	-3.25	0.001	.9520808	.98789
pyaccrati~2		1.033874	.1137436	0.30	0.762	.8333379	1.282668
pyperstudlep		.9986489	.0073992	-0.18	0.855	.9842515	1.013257
pyperstude~s		.9943725	.0070675	-0.79	0.427	.9806164	1.008322
pyperstudw~e		.9651599	.0170483	-2.01	0.045	.9323177	.9991591
pyperstudb~k		.9711439	.020947	-1.36	0.175	.9309442	1.013079
pyperstudl~o		.9751855	.0184926	-1.33	0.185	.9396059	1.012112
pyperstudm~e		1.001422	.0090163	0.16	0.875	.9839051	1.01925
Asian		5.398266	6.105324	1.49	0.136	.5882519	49.53877
Black		1.6798	.4440033	1.96	0.050	1.000619	2.819982
Latino		2.357272	.5476327	3.69	0.000	1.49507	3.716701
Other		3.678121	4.156841	1.15	0.249	.4014559	33.69879
AsianMale		.1138144	.2078195	-1.19	0.234	.0031765	4.078015
BlackMale		1.129462	.559332	0.25	0.806	.4278977	2.981285
LatinoMale		1.1296	.4395339	0.31	0.754	.5268865	2.421766
OtherMale		.2756537	.370586	-0.96	0.338	.0197705	3.843355
-----							
3							
tenure		1.028126	.010148	2.81	0.005	1.008427	1.048209
exper		1.030723	.0121203	2.57	0.010	1.00724	1.054754
male		.9010956	.1494215	-0.63	0.530	.6510629	1.24715
age		.9818607	.0103245	-1.74	0.082	.9618322	1.002306
pyaccrati~2		1.449102	.1702673	3.16	0.002	1.151023	1.824374
pyperstudlep		.9940002	.0077828	-0.77	0.442	.9788626	1.009372
pyperstude~s		1.018039	.0074584	2.44	0.015	1.003525	1.032762
pyperstudw~e		.9731626	.0188348	-1.41	0.160	.9369384	1.010787
pyperstudb~k		.9767193	.0230213	-1.00	0.318	.9326248	1.022899
pyperstudl~o		.9719886	.0200612	-1.38	0.169	.933454	1.012114
pyperstudm~e		.9783682	.0115252	-1.86	0.063	.9560379	1.00122
Asian		3.914588	4.554964	1.17	0.241	.4001649	38.29422
Black		.4016152	.1464152	-2.50	0.012	.1965581	.8205958
Latino		1.249375	.3197785	0.87	0.384	.7565318	2.063282
Other		4.498746	5.258698	1.29	0.198	.4550934	44.47157
AsianMale		3.05e-06	.0013437	-0.03	0.977	0	.
BlackMale		3.140304	1.92546	1.87	0.062	.9441961	10.44434
LatinoMale		2.032743	.8632377	1.67	0.095	.8843127	4.672606
OtherMale		.1936018	.2833333	-1.12	0.262	.0109948	3.409026
-----							

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -255.93437  
Iteration 1: log likelihood = -229.85445  
Iteration 2: log likelihood = -226.97003  
Iteration 3: log likelihood = -226.33104  
Iteration 4: log likelihood = -226.29046  
Iteration 5: log likelihood = -226.28355  
Iteration 6: log likelihood = -226.28197  
Iteration 7: log likelihood = -226.2816  
Iteration 8: log likelihood = -226.28152  
Iteration 9: log likelihood = -226.28151

Multinomial logistic regression	Number of obs	=	257
	LR chi2(32)	=	59.31
	Prob > chi2	=	0.0023
Log likelihood = -226.28151	Pseudo R2	=	0.1159

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						

tenure		1.01741	.0316839	0.55	0.579	.957168	1.081444
exper		1.008852	.0405751	0.22	0.827	.9323802	1.091596
male		1.562747	.7259933	0.96	0.337	.6287129	3.884408
age		1.002999	.0365796	0.08	0.935	.9338071	1.077319
pyacctrati~2		.8270188	.3276385	-0.48	0.632	.3804491	1.797771
pyperstudlep		1.078152	.1026485	0.79	0.429	.8946203	1.299334
pyperstude~s		.9909184	.0196508	-0.46	0.645	.9531425	1.030192
pyperstudw~e		.8325945	.2427783	-0.63	0.530	.4701425	1.474475
pyperstudb~k		.7969001	.2501295	-0.72	0.469	.430755	1.474271
pyperstudl~o		.8468617	.243888	-0.58	0.564	.4815869	1.489191
pyperstudm~e		1.00972	.0314572	0.31	0.756	.9499093	1.073296
Asian		(omitted)					
Black		1.75e+07	2.72e+10	0.01	0.991	0	.
Latino		2.83e-06	.0032815	-0.01	0.991	0	.
Other		4.030442	6.199712	0.91	0.365	.1977044	82.16541
AsianMale		(omitted)					
BlackMale		1.03e-07	.0003819	-0.00	0.997	0	.
LatinoMale		4035030	4.68e+09	0.01	0.990	0	.
OtherMale		(omitted)					
-----							
3							
tenure		1.003098	.0216289	0.14	0.886	.9615894	1.046398
exper		1.044263	.0299574	1.51	0.131	.9871674	1.10466
male		2.662615	.8791991	2.97	0.003	1.393929	5.085999
age		.9949266	.0253699	-0.20	0.842	.9464246	1.045914
pyacctrati~2		1.479697	.3994219	1.45	0.147	.8717734	2.51155
pyperstudlep		1.117561	.0722803	1.72	0.086	.984506	1.268599
pyperstude~s		1.017879	.0133167	1.35	0.176	.9921107	1.044317
pyperstudw~e		1.037453	.211892	0.18	0.857	.6952118	1.548175
pyperstudb~k		.9721206	.2127067	-0.13	0.897	.6330987	1.492687
pyperstudl~o		1.046436	.2110404	0.23	0.822	.7047678	1.553742
pyperstudm~e		.9848224	.0250647	-0.60	0.548	.9369016	1.035194
Asian		(omitted)					
Black		7429339	1.16e+10	0.01	0.992	0	.
Latino		3.756357	4.66192	1.07	0.286	.3298846	42.77319
Other		.9164823	1.433656	-0.06	0.956	.0427157	19.6635
AsianMale		(omitted)					
BlackMale		.4620364	1038.481	-0.00	1.000	0	.
LatinoMale		.496143	.7733054	-0.45	0.653	.0233827	10.52734
OtherMale		(omitted)					
-----							

-----> region = 15

note: Asian omitted because of collinearity  
note: Other omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -286.86545  
Iteration 1: log likelihood = -263.30984  
Iteration 2: log likelihood = -261.77967  
Iteration 3: log likelihood = -261.75364  
Iteration 4: log likelihood = -261.75009  
Iteration 5: log likelihood = -261.74947  
Iteration 6: log likelihood = -261.74933  
Iteration 7: log likelihood = -261.74929  
Iteration 8: log likelihood = -261.74929

Multinomial logistic regression	Number of obs	=	276
	LR chi2(30)	=	50.23
	Prob > chi2	=	0.0117
Log likelihood = -261.74929	Pseudo R2	=	0.0876

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						

1		(base outcome)					
2							
	tenure	1.066515	.0364706	1.88	0.060	.9973763	1.140446
	exper	.9319033	.0336131	-1.96	0.051	.8682977	1.000168
	male	1.415279	.6546325	0.75	0.453	.5716387	3.503987
	age	1.015902	.0265234	0.60	0.546	.9652252	1.06924
	pyacctrati~2	.5813986	.2064462	-1.53	0.127	.2898843	1.166066
	pyperstudlep	.9681011	.0337911	-0.93	0.353	.9040864	1.036649
	pyperstude~s	1.013429	.016424	0.82	0.410	.9817446	1.046136
	pyperstudw~e	.8682212	.2343689	-0.52	0.601	.511512	1.473686
	pyperstudb~k	.9257482	.264932	-0.27	0.787	.5283172	1.62215
	pyperstudl~o	.8649255	.2351371	-0.53	0.593	.5076596	1.473618
	pyperstudm~e	.980838	.0313746	-0.60	0.545	.921233	1.0443
	Asian	(omitted)					
	Black	1.07e+07	1.31e+10	0.01	0.989	0	.
	Latino	3.550699	2.07576	2.17	0.030	1.129006	11.16687
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	5.49e-07	.0006731	-0.01	0.991	0	.
	LatinoMale	.8075745	.6599843	-0.26	0.794	.1627588	4.007013
	OtherMale	(omitted)					
3							
	tenure	.9741226	.0191349	-1.33	0.182	.9373316	1.012358
	exper	1.048028	.0257425	1.91	0.056	.9987686	1.099716
	male	.9040142	.2941738	-0.31	0.756	.4777351	1.710659
	age	.9953251	.0224992	-0.21	0.836	.9521901	1.040414
	pyacctrati~2	1.289076	.2964874	1.10	0.270	.8213003	2.023274
	pyperstudlep	1.009905	.0286073	0.35	0.728	.9553635	1.06756
	pyperstude~s	1.016344	.0122945	1.34	0.180	.9925306	1.040729
	pyperstudw~e	.9973852	.1950462	-0.01	0.989	.6798358	1.463261
	pyperstudb~k	1.039757	.2171388	0.19	0.852	.6905102	1.565646
	pyperstudl~o	.988651	.1939818	-0.06	0.954	.6730231	1.452299
	pyperstudm~e	.9714734	.0249477	-1.13	0.260	.9237869	1.021621
	Asian	(omitted)					
	Black	1.385148	2516.989	0.00	1.000	0	.
	Latino	.8026128	.4260471	-0.41	0.679	.2835743	2.27167
	Other	(omitted)					
	AsianMale	(omitted)					
	BlackMale	1.024447	1861.55	0.00	1.000	0	.
	LatinoMale	1.840172	1.414833	0.79	0.428	.4077598	8.304478
	OtherMale	(omitted)					

-----> region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -534.44334  
Iteration 1: log likelihood = -494.76016  
Iteration 2: log likelihood = -493.57752  
Iteration 3: log likelihood = -493.498  
Iteration 4: log likelihood = -493.48575  
Iteration 5: log likelihood = -493.48279  
Iteration 6: log likelihood = -493.48216  
Iteration 7: log likelihood = -493.48202  
Iteration 8: log likelihood = -493.48199

Multinomial logistic regression	Number of obs	=	492
	LR chi2(34)	=	81.92
	Prob > chi2	=	0.0000
Log likelihood = -493.48199	Pseudo R2	=	0.0766

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-------	--	-----	-----------	---	------	----------------------

1		(base outcome)					
2							
	tenure	1.044386	.0212426	2.14	0.033	1.003571	1.086862
	exper	.958234	.0244738	-1.67	0.095	.911447	1.007423
	male	.8817168	.232552	-0.48	0.633	.5258059	1.478539
	age	.9721568	.0201153	-1.36	0.172	.9335202	1.012392
	pyacctrati~2	.9422033	.1898865	-0.30	0.768	.6347433	1.398592
	pyperstudlep	1.016007	.0166476	0.97	0.332	.9838962	1.049165
	pyperstude~s	.9821423	.0085544	-2.07	0.039	.9655183	.9990526
	pyperstudw~e	.9450086	.0338828	-1.58	0.115	.8808793	1.013807
	pyperstudb~k	.9714633	.0391886	-0.72	0.473	.897613	1.051389
	pyperstudl~o	.9542162	.033311	-1.34	0.179	.8911112	1.02179
	pyperstudm~e	.9998232	.0105669	-0.02	0.987	.9793255	1.02075
	Asian	(omitted)					
	Black	.7135945	.7228972	-0.33	0.739	.0979836	5.196963
	Latino	2.476497	1.397719	1.61	0.108	.8192665	7.486011
	Other	1.685438	1.812926	0.49	0.627	.2047037	13.87714
	AsianMale	(omitted)					
	BlackMale	.000013	.0056439	-0.03	0.979	0	.
	LatinoMale	.6681335	.6491099	-0.42	0.678	.0995171	4.485685
	OtherMale	3280409	2.64e+09	0.02	0.985	0	.
3							
	tenure	1.028578	.0168474	1.72	0.085	.9960823	1.062134
	exper	.9974883	.0205616	-0.12	0.903	.9579914	1.038614
	male	2.112691	.5205149	3.04	0.002	1.30353	3.424137
	age	1.006296	.0191373	0.33	0.741	.9694777	1.044512
	pyacctrati~2	1.444244	.2665976	1.99	0.046	1.005807	2.073799
	pyperstudlep	1.027443	.015542	1.79	0.074	.9974278	1.05836
	pyperstude~s	1.009684	.0075053	1.30	0.195	.9950807	1.024502
	pyperstudw~e	1.029177	.0443259	0.67	0.504	.9458658	1.119827
	pyperstudb~k	1.002614	.0485633	0.05	0.957	.91181	1.102461
	pyperstudl~o	1.015563	.042965	0.37	0.715	.9347495	1.103362
	pyperstudm~e	.9933764	.0119202	-0.55	0.580	.9702859	1.017016
	Asian	(omitted)					
	Black	1.334103	1.303816	0.29	0.768	.1964771	9.058717
	Latino	1.397314	.9005253	0.52	0.604	.3951129	4.941592
	Other	.8155557	1.02043	-0.16	0.871	.0702139	9.472928
	AsianMale	(omitted)					
	BlackMale	.6134384	1.085177	-0.28	0.782	.0191421	19.65862
	LatinoMale	1.01834	1.031909	0.02	0.986	.1397491	7.420557
	OtherMale	.6685122	767.0158	-0.00	1.000	0	.

-----> region = 17

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
Iteration 0: log likelihood = -316.58585  
Iteration 1: log likelihood = -275.99917  
Iteration 2: log likelihood = -273.68903  
Iteration 3: log likelihood = -273.59325  
Iteration 4: log likelihood = -273.57625  
Iteration 5: log likelihood = -273.57398  
Iteration 6: log likelihood = -273.57368  
Iteration 7: log likelihood = -273.57363  
Iteration 8: log likelihood = -273.57361

Multinomial logistic regression

Number of obs = 311  
LR chi2(32) = 86.02  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.1359

Log likelihood = -273.57361

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9788785	.0292408	-0.71	0.475	.9232129	1.0379
	exper	1.060777	.0431739	1.45	0.147	.979445	1.148863
	male	.9210065	.3730605	-0.20	0.839	.4163678	2.037268
	age	.9203789	.0347493	-2.20	0.028	.8547304	.9910697
pyacctrati~2		.9627456	.2903794	-0.13	0.900	.5330594	1.738791
pyperstudlep		.9414593	.0427958	-1.33	0.184	.8612091	1.029187
pyperstude~s		.9928478	.0177577	-0.40	0.688	.9586463	1.028269
pyperstudw~e		1.032933	.1242707	0.27	0.788	.8159533	1.307612
pyperstudb~k		1.234801	.1882209	1.38	0.166	.9159001	1.664737
pyperstudl~o		1.020968	.1193356	0.18	0.859	.8119319	1.283821
pyperstudm~e		.9523447	.0316999	-1.47	0.142	.8921973	1.016547
	Asian	(omitted)					
	Black	3.51e+07	1.14e+11	0.01	0.996	0	.
	Latino	4.02853	3.598798	1.56	0.119	.6994308	23.20323
	Other	.0000226	.033569	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.3163206	.4052207	-0.90	0.369	.0256862	3.895421
	OtherMale	.029042	61.00393	-0.00	0.999	0	.
3							
	tenure	1.011061	.0207986	0.53	0.593	.9711077	1.052659
	exper	1.01528	.0251991	0.61	0.541	.9670731	1.065891
	male	2.986319	.9060417	3.61	0.000	1.647721	5.412384
	age	.9892668	.0223641	-0.48	0.633	.9463909	1.034085
pyacctrati~2		1.37428	.3165315	1.38	0.167	.8750285	2.158383
pyperstudlep		.9776457	.0247488	-0.89	0.372	.9303226	1.027376
pyperstude~s		1.042218	.0129651	3.32	0.001	1.017114	1.067941
pyperstudw~e		1.101733	.1326434	0.80	0.421	.8701535	1.394945
pyperstudb~k		1.089762	.1548163	0.61	0.545	.824909	1.43965
pyperstudl~o		1.101571	.1312526	0.81	0.417	.8721507	1.391341
pyperstudm~e		.9981324	.0140476	-0.13	0.894	.9709759	1.026049
	Asian	(omitted)					
	Black	1.349456	6181.668	0.00	1.000	0	.
	Latino	.2579871	.3038912	-1.15	0.250	.0256414	2.595704
	Other	1.34e-06	.0012567	-0.01	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	4.57711	6.331161	1.10	0.271	.3042218	68.86402
	OtherMale	1.225541	1624.82	0.00	1.000	0	.

-----> region = 18

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -433.29228  
Iteration 1: log likelihood = -390.38619  
Iteration 2: log likelihood = -388.40657  
Iteration 3: log likelihood = -388.33147  
Iteration 4: log likelihood = -388.31625  
Iteration 5: log likelihood = -388.31303  
Iteration 6: log likelihood = -388.31232  
Iteration 7: log likelihood = -388.31214  
Iteration 8: log likelihood = -388.31211  
Iteration 9: log likelihood = -388.3121

Multinomial logistic regression	Number of obs	=	407
	LR chi2(34)	=	89.96
	Prob > chi2	=	0.0000

Log likelihood = -388.3121 Pseudo R2 = 0.1038

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.011878	.0218223	0.55	0.584	.9699985	1.055566
exper	.9934203	.0261813	-0.25	0.802	.9434087	1.046083
male	1.057789	.3704449	0.16	0.873	.5324787	2.101336
age	.9671675	.0204463	-1.58	0.114	.9279124	1.008083
pyacctrati~2	.6165529	.1690084	-1.76	0.078	.3602803	1.055116
pyperstudlep	1.008917	.0164648	0.54	0.586	.9771572	1.041709
pyperstude~s	.9990209	.0134206	-0.07	0.942	.9730603	1.025674
pyperstudw~e	.7027714	.1294871	-1.91	0.056	.4897544	1.008439
pyperstudb~k	.7444857	.1432346	-1.53	0.125	.5106111	1.085482
pyperstudl~o	.7043056	.12904	-1.91	0.056	.4918215	1.00859
pyperstudm~e	1.006977	.0126798	0.55	0.581	.9824293	1.032138
Asian	(omitted)					
Black	7.90329	9.267513	1.76	0.078	.7937371	78.69356
Latino	2.534413	1.084328	2.17	0.030	1.095716	5.86215
Other	.5455238	1650.421	-0.00	1.000	0	.
AsianMale	(omitted)					
BlackMale	.8111528	1.362771	-0.12	0.901	.0301334	21.83522
LatinoMale	.9303287	.5451168	-0.12	0.902	.2950417	2.933523
OtherMale	3172368	9.94e+09	0.00	0.996	0	.
3						
tenure	1.013158	.0175596	0.75	0.451	.9793201	1.048166
exper	1.053478	.0261924	2.10	0.036	1.003372	1.106085
male	1.394962	.4107798	1.13	0.258	.7832594	2.484385
age	.9602142	.020427	-1.91	0.056	.9210012	1.001097
pyacctrati~2	1.425607	.3278295	1.54	0.123	.9083625	2.237383
pyperstudlep	.9863537	.0149508	-0.91	0.365	.9574817	1.016096
pyperstude~s	1.044195	.0135009	3.34	0.001	1.018066	1.070994
pyperstudw~e	.8682601	.1523887	-0.80	0.421	.6155384	1.224742
pyperstudb~k	.9000217	.1657834	-0.57	0.567	.6272812	1.291349
pyperstudl~o	.8561932	.1504489	-0.88	0.377	.6067364	1.208213
pyperstudm~e	.9885997	.0143263	-0.79	0.429	.9609157	1.017081
Asian	(omitted)					
Black	5.620137	6.884263	1.41	0.159	.5094322	62.00225
Latino	.8693558	.3625682	-0.34	0.737	.3838846	1.968768
Other	1213388	2.16e+09	0.01	0.994	0	.
AsianMale	(omitted)					
BlackMale	.2607717	.4955494	-0.71	0.479	.0062908	10.80973
LatinoMale	.6891442	.4131079	-0.62	0.535	.2128417	2.231328
OtherMale	1.661754	3258.332	0.00	1.000	0	.

-----> region = 19

Iteration 0: log likelihood = -875.07174  
Iteration 1: log likelihood = -804.47026  
Iteration 2: log likelihood = -800.46546  
Iteration 3: log likelihood = -800.02137  
Iteration 4: log likelihood = -799.95459  
Iteration 5: log likelihood = -799.93848  
Iteration 6: log likelihood = -799.93506  
Iteration 7: log likelihood = -799.93435  
Iteration 8: log likelihood = -799.93418  
Iteration 9: log likelihood = -799.93415  
Iteration 10: log likelihood = -799.93414

Multinomial logistic regression

Number of obs = 815  
LR chi2(38) = 150.28

Log likelihood = -799.93414

Prob > chi2 = 0.0000  
Pseudo R2 = 0.0859

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.015348	.0143939	1.07	0.283	.9875244	1.043955
	exper	1.023392	.0172807	1.37	0.171	.9900767	1.057828
	male	.9345223	.2816677	-0.22	0.822	.5176486	1.687113
	age	.9865632	.0127207	-1.05	0.294	.9619435	1.011813
pyacctrati~2		1.008378	.1614764	0.05	0.958	.7367439	1.380163
pyperstudlep		1.007356	.006267	1.18	0.239	.9951474	1.019714
pyperstude~s		.9827135	.0091038	-1.88	0.060	.9650315	1.00072
pyperstudw~e		.9281865	.0940467	-0.74	0.462	.7610074	1.132092
pyperstudb~k		1.03772	.1044303	0.37	0.713	.8519619	1.263981
pyperstudl~o		.9787923	.0918388	-0.23	0.819	.8143731	1.176407
pyperstudm~e		.9977976	.0134355	-0.16	0.870	.971809	1.024481
	Asian	6.69e-08	.0002474	-0.00	0.996	0	.
	Black	.4459049	.2489425	-1.45	0.148	.1492892	1.331853
	Latino	1.192716	.276115	0.76	0.446	.7576756	1.877548
	Other	.9859851	.9361161	-0.01	0.988	.1533631	6.338988
	AsianMale	8693180	8.26e+10	0.00	0.999	0	.
	BlackMale	6.017317	5.73131	1.88	0.060	.9303783	38.91762
	LatinoMale	1.156453	.4263217	0.39	0.693	.5614821	2.381881
	OtherMale	3.65e-07	.0006261	-0.01	0.993	0	.
3							
	tenure	1.021414	.0152527	1.42	0.156	.9919527	1.051751
	exper	1.095227	.0219941	4.53	0.000	1.052957	1.139194
	male	1.456241	.4961388	1.10	0.270	.7468448	2.839463
	age	.9880129	.0163251	-0.73	0.465	.9565289	1.020533
pyacctrati~2		1.190261	.2207646	0.94	0.348	.8274938	1.712062
pyperstudlep		1.022046	.0073082	3.05	0.002	1.007822	1.036471
pyperstude~s		1.016694	.0123499	1.36	0.173	.9927745	1.04119
pyperstudw~e		.8851958	.0962462	-1.12	0.262	.7153018	1.095442
pyperstudb~k		.9285612	.102655	-0.67	0.503	.7476663	1.153223
pyperstudl~o		.889556	.0898484	-1.16	0.247	.7297916	1.084296
pyperstudm~e		.9737601	.0199105	-1.30	0.193	.9355078	1.013577
	Asian	3.17e-07	.0014253	-0.00	0.997	0	.
	Black	.6905427	.4423999	-0.58	0.563	.1967269	2.423915
	Latino	1.041449	.2940732	0.14	0.886	.5988046	1.811303
	Other	3.09e-07	.0004188	-0.01	0.991	0	.
	AsianMale	1.82e+13	1.35e+17	0.00	0.997	0	.
	BlackMale	7.19e-07	.0008535	-0.01	0.990	0	.
	LatinoMale	1.174408	.4991673	0.38	0.705	.5105363	2.701541
	OtherMale	1.14e+07	1.54e+10	0.01	0.990	0	.

-----> region = 20

Iteration 0: log likelihood = -2100.9698  
Iteration 1: log likelihood = -1997.1491  
Iteration 2: log likelihood = -1993.0417  
Iteration 3: log likelihood = -1992.6693  
Iteration 4: log likelihood = -1992.6095  
Iteration 5: log likelihood = -1992.5955  
Iteration 6: log likelihood = -1992.5923  
Iteration 7: log likelihood = -1992.5916  
Iteration 8: log likelihood = -1992.5915  
Iteration 9: log likelihood = -1992.5915

Multinomial logistic regression

Number of obs = 2032  
LR chi2(38) = 216.76



Log likelihood = -1992.5915      Prob > chi2 = 0.0000  
Pseudo R2 = 0.0516

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9988131	.008232	-0.14	0.885	.9828082	1.015079
	exper	1.000565	.0102856	0.05	0.956	.9806075	1.020929
	male	1.238481	.1917286	1.38	0.167	.9143563	1.677504
	age	.9808584	.0084939	-2.23	0.026	.9643512	.9976483
pyacctrati~2		1.06565	.100697	0.67	0.501	.8854854	1.282471
pyperstudlep		.9836721	.0066094	-2.45	0.014	.9708028	.996712
pyperstude~s		1.00391	.0056602	0.69	0.489	.9928772	1.015065
pyperstudw~e		1.000016	.0349781	0.00	1.000	.9337577	1.070977
pyperstudb~k		.9826733	.0371762	-0.46	0.644	.9124452	1.058307
pyperstudl~o		.9940472	.0358907	-0.17	0.869	.926134	1.06694
pyperstudm~e		1.00513	.0064055	0.80	0.422	.9926531	1.017763
	Asian	.6744496	.7846726	-0.34	0.735	.0689666	6.595688
	Black	1.704299	.5003289	1.82	0.069	.9586486	3.029925
	Latino	1.212287	.1830156	1.28	0.202	.9017823	1.629704
	Other	.3375857	.3672657	-1.00	0.318	.0400264	2.847226
	AsianMale	5.12e-07	.0007836	-0.01	0.992	0	.
	BlackMale	1.63494	.7248291	1.11	0.267	.6856978	3.89826
	LatinoMale	1.189458	.2756556	0.75	0.454	.7552382	1.873329
	OtherMale	4.908075	7.284675	1.07	0.284	.2676325	90.0085
3							
	tenure	1.024231	.0084384	2.91	0.004	1.007825	1.040904
	exper	1.074677	.0140183	5.52	0.000	1.04755	1.102506
	male	1.592707	.2712603	2.73	0.006	1.140681	2.22386
	age	.954898	.0112262	-3.93	0.000	.9331466	.9771565
pyacctrati~2		1.370749	.1440548	3.00	0.003	1.115587	1.684273
pyperstudlep		1.008042	.0065327	1.24	0.216	.9953189	1.020927
pyperstude~s		1.028462	.006597	4.38	0.000	1.015613	1.041473
pyperstudw~e		1.033827	.0400311	0.86	0.390	.9582712	1.115341
pyperstudb~k		1.015695	.0423157	0.37	0.709	.9360534	1.102112
pyperstudl~o		1.00334	.0400892	0.08	0.933	.9277642	1.085072
pyperstudm~e		.9858113	.0097609	-1.44	0.149	.9668647	1.005129
	Asian	4.20e-07	.0005776	-0.01	0.991	0	.
	Black	1.862858	.5668195	2.04	0.041	1.026079	3.382037
	Latino	1.104792	.1843224	0.60	0.550	.7966487	1.532125
	Other	3.87e-07	.0004376	-0.01	0.990	0	.
	AsianMale	3178031	4.37e+09	0.01	0.991	0	.
	BlackMale	.6770305	.3702645	-0.71	0.476	.2317878	1.977543
	LatinoMale	.8371366	.2206588	-0.67	0.500	.4993789	1.403339
	OtherMale	1.761016	3221.052	0.00	1.000	0	.

by pyschooltype:

```
. bysort pyschooltype : mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2007a==1, rrr iter(20)
```

```
-----> pyschooltype =
no observations
```

```
-----> pyschooltype = B
```

note: Asian omitted because of collinearity

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -335.10713
Iteration 1: log likelihood = -311.05446
Iteration 2: log likelihood = -303.19194
Iteration 3: log likelihood = -302.8472
Iteration 4: log likelihood = -302.82519
Iteration 5: log likelihood = -302.82161
Iteration 6: log likelihood = -302.82076
Iteration 7: log likelihood = -302.82057
Iteration 8: log likelihood = -302.82053
Iteration 9: log likelihood = -302.82052
```

```
Multinomial logistic regression      Number of obs   =      375
                                     LR chi2(34)        =      64.57
                                     Prob > chi2         =      0.0012
Log likelihood = -302.82052          Pseudo R2        =      0.0963
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.025216	.0327692	0.78	0.436	.9629599	1.091497
	exper	1.020172	.0417481	0.49	0.626	.9415424	1.105368
	male	1.816421	.924234	1.17	0.241	.6700495	4.924092
	age	.9520422	.033192	-1.41	0.159	.88916	1.019371
pyacctrati~2		.3822308	.1470166	-2.50	0.012	.179858	.8123094
pyperstudlep		1.003793	.031354	0.12	0.904	.9441839	1.067166
pyperstude~s		.9595152	.0176632	-2.25	0.025	.925513	.9947666
pyperstudw~e		.9396728	.032343	-1.81	0.071	.8783726	1.005251
pyperstudb~k		.9859194	.0381074	-0.37	0.714	.9139892	1.063511
pyperstudl~o		.9746911	.0357023	-0.70	0.484	.9071687	1.047239
pyperstudm~e		.9779775	.030562	-0.71	0.476	.9198747	1.03975
	Asian	(omitted)					
	Black	4.263881	4.544183	1.36	0.174	.5280153	34.43211
	Latino	5.507124	5.114491	1.84	0.066	.8920931	33.99692
	Other	1.36e-06	.0044298	-0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	.2351573	.3207763	-1.06	0.289	.0162271	3.407815
	LatinoMale	.7635596	1.024608	-0.20	0.841	.0550337	10.59392
	OtherMale	.6337564	2927.853	-0.00	1.000	0	.
3							
	tenure	.9659513	.0163066	-2.05	0.040	.9345138	.9984463
	exper	1.017182	.0192445	0.90	0.368	.9801543	1.055609
	male	.8408847	.2127931	-0.68	0.493	.5120739	1.38083
	age	1.016265	.017854	0.92	0.358	.9818673	1.051868
pyacctrati~2		.7688479	.1489357	-1.36	0.175	.5259588	1.123904
pyperstudlep		.9690469	.018641	-1.63	0.102	.9331915	1.00628
pyperstude~s		.9965173	.0088685	-0.39	0.695	.9792862	1.014052
pyperstudw~e		.9708822	.0243543	-1.18	0.239	.9243031	1.019808
pyperstudb~k		.9684587	.0264746	-1.17	0.241	.9179349	1.021763
pyperstudl~o		.9760611	.0251794	-0.94	0.348	.9279373	1.026681
pyperstudm~e		.9946528	.0161975	-0.33	0.742	.9634075	1.026911
	Asian	(omitted)					
	Black	2.209132	2.07615	0.84	0.399	.3501543	13.93747
	Latino	1.089026	.8357317	0.11	0.912	.2420005	4.900727
	Other	5.51e-07	.0009326	-0.01	0.993	0	.
	AsianMale	(omitted)					
	BlackMale	1.03381	1.129095	0.03	0.976	.1215582	8.792187
	LatinoMale	5.519606	6.341227	1.49	0.137	.5807657	52.45842
	OtherMale	.9906558	2370.578	-0.00	1.000	0	.

-----> pyschooltype = E

```

Iteration 0: log likelihood = -11371.741
Iteration 1: log likelihood = -10756.6
Iteration 2: log likelihood = -10751.222
Iteration 3: log likelihood = -10751.216
Iteration 4: log likelihood = -10751.216

```

```

Multinomial logistic regression
Number of obs   =      10448
LR chi2(38)     =      1241.05
Prob > chi2     =      0.0000
Pseudo R2       =      0.0546

Log likelihood = -10751.216

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.020283	.0042902	4.78	0.000	1.011909	1.028727
	exper	1.006411	.0052792	1.22	0.223	.9961169	1.016812
	male	1.671669	.1440633	5.96	0.000	1.411869	1.979276
	age	.9744186	.0040043	-6.31	0.000	.9666018	.9822986
pyacctrati~2		.9027747	.0378591	-2.44	0.015	.8315398	.9801121
pyperstudlep		1.007944	.0018502	4.31	0.000	1.004324	1.011577
pyperstude~s		.9885757	.0021323	-5.33	0.000	.9844052	.9927637
pyperstudw~e		.9972726	.0053109	-0.51	0.608	.9869176	1.007736
pyperstudb~k		.9977568	.0060247	-0.37	0.710	.9860183	1.009635
pyperstudl~o		1.000259	.0056065	0.05	0.963	.9893305	1.011308
pyperstudm~e		1.010942	.0039411	2.79	0.005	1.003248	1.018696
	Asian	.7038376	.289238	-0.85	0.393	.3145401	1.574957
	Black	1.009211	.0960074	0.10	0.923	.837541	1.216067
	Latino	1.172664	.0906094	2.06	0.039	1.007866	1.364408
	Other	1.101515	.2973064	0.36	0.720	.6490006	1.869543
	AsianMale	1.650969	1.308569	0.63	0.527	.349197	7.805614
	BlackMale	.9660861	.1958915	-0.17	0.865	.649261	1.437515
	LatinoMale	.6415602	.0938302	-3.03	0.002	.4816666	.8545318
	OtherMale	1.479073	.9457259	0.61	0.540	.4224037	5.179068
3							
	tenure	1.021637	.0036651	5.97	0.000	1.014479	1.028846
	exper	1.062944	.0052105	12.45	0.000	1.052781	1.073206
	male	2.610026	.2030789	12.33	0.000	2.240862	3.040006
	age	.9857676	.0040552	-3.48	0.000	.9778515	.9937479
pyacctrati~2		.9612653	.0391413	-0.97	0.332	.8875312	1.041125
pyperstudlep		.9969375	.001804	-1.70	0.090	.9934081	1.00048
pyperstude~s		1.00546	.0020625	2.65	0.008	1.001425	1.00951
pyperstudw~e		1.006002	.0054997	1.09	0.274	.9952802	1.016839
pyperstudb~k		.9946084	.0061254	-0.88	0.380	.9826749	1.006687
pyperstudl~o		.9973552	.0057306	-0.46	0.645	.9861865	1.00865
pyperstudm~e		.9987185	.0039671	-0.32	0.747	.9909733	1.006524
	Asian	1.181207	.4567534	0.43	0.667	.5535798	2.520413
	Black	.923678	.0890007	-0.82	0.410	.7647216	1.115676
	Latino	1.16524	.0902829	1.97	0.048	1.001069	1.356333
	Other	.9190538	.2667562	-0.29	0.771	.5203301	1.623315
	AsianMale	1.092666	.8169692	0.12	0.906	.2523816	4.730613
	BlackMale	.588156	.1202923	-2.60	0.009	.3939136	.8781812
	LatinoMale	.569616	.0796001	-4.03	0.000	.433144	.7490866
	OtherMale	.1874132	.1736784	-1.81	0.071	.0304775	1.152446

```

-----> pyschooltype = M

```

```

Iteration 0: log likelihood = -6203.2011
Iteration 1: log likelihood = -5921.1494
Iteration 2: log likelihood = -5916.2875
Iteration 3: log likelihood = -5916.275

```

Iteration 4: log likelihood = -5916.275

Multinomial logistic regression

Number of obs = 5858  
 LR chi2(38) = 573.85  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0463

Log likelihood = -5916.275

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.01126	.0050139	2.26	0.024	1.001481	1.021135
	exper	1.004477	.0061677	0.73	0.467	.992461	1.016638
	male	1.605584	.1271111	5.98	0.000	1.374817	1.875085
	age	.9757079	.0048335	-4.96	0.000	.9662802	.9852275
pyacctrati~2		.8885108	.0502436	-2.09	0.037	.7952961	.9926508
pyperstudlep		1.008298	.0036491	2.28	0.022	1.001171	1.015476
pyperstude~s		.9956417	.0028473	-1.53	0.127	.9900767	1.001238
pyperstudw~e		.9943037	.0074603	-0.76	0.446	.9797888	1.009034
pyperstudb~k		.9909908	.0084195	-1.07	0.287	.9746255	1.007631
pyperstudl~o		.991883	.0080169	-1.01	0.313	.976294	1.007721
pyperstudm~e		1.016332	.0050713	3.25	0.001	1.00644	1.02632
	Asian	1.562366	.6367442	1.09	0.274	.7028696	3.47289
	Black	.9701946	.1093059	-0.27	0.788	.7779634	1.209925
	Latino	1.874829	.2181182	5.40	0.000	1.492563	2.354999
	Other	1.092287	.4060456	0.24	0.812	.5271232	2.263401
	AsianMale	.6755661	.4383992	-0.60	0.546	.1893619	2.410145
	BlackMale	1.519711	.2710525	2.35	0.019	1.071377	2.155656
	LatinoMale	.8259046	.1310454	-1.21	0.228	.6051606	1.127169
	OtherMale	1.647611	.8924733	0.92	0.357	.5698732	4.763554
3							
	tenure	1.015052	.0052116	2.91	0.004	1.004889	1.025318
	exper	1.065929	.0078236	8.70	0.000	1.050705	1.081374
	male	2.028776	.1792973	8.00	0.000	1.706112	2.412464
	age	.9642353	.006181	-5.68	0.000	.9521966	.9764263
pyacctrati~2		.9063547	.0605362	-1.47	0.141	.795144	1.03312
pyperstudlep		.9948109	.0046673	-1.11	0.267	.9857051	1.004001
pyperstude~s		1.006991	.003329	2.11	0.035	1.000488	1.013537
pyperstudw~e		1.023489	.010511	2.26	0.024	1.003094	1.044299
pyperstudb~k		1.003257	.0115629	0.28	0.778	.9808482	1.026178
pyperstudl~o		1.009274	.0110409	0.84	0.399	.9878642	1.031147
pyperstudm~e		1.013876	.0059831	2.34	0.020	1.002216	1.02567
	Asian	.8022219	.5228945	-0.34	0.735	.2236014	2.878156
	Black	.8259408	.1239934	-1.27	0.203	.6154081	1.108497
	Latino	1.519405	.221136	2.87	0.004	1.142322	2.020963
	Other	.7313265	.4087408	-0.56	0.576	.2445525	2.187009
	AsianMale	.6760348	.6978436	-0.38	0.704	.0893939	5.112465
	BlackMale	1.822302	.3934864	2.78	0.005	1.193498	2.782394
	LatinoMale	.8989786	.1699392	-0.56	0.573	.6206427	1.302138
	OtherMale	1.430376	1.069508	0.48	0.632	.3303678	6.193027

-----> pyschooltype = S

Iteration 0: log likelihood = -6925.5249  
 Iteration 1: log likelihood = -6628.0653  
 Iteration 2: log likelihood = -6616.8508  
 Iteration 3: log likelihood = -6616.6579  
 Iteration 4: log likelihood = -6616.6171  
 Iteration 5: log likelihood = -6616.6075  
 Iteration 6: log likelihood = -6616.6053  
 Iteration 7: log likelihood = -6616.6049  
 Iteration 8: log likelihood = -6616.6048

Iteration 9: log likelihood = -6616.6048

Multinomial logistic regression	Number of obs	=	7193
	LR chi2(38)	=	617.84
	Prob > chi2	=	0.0000
Log likelihood = -6616.6048	Pseudo R2	=	0.0446

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.006243	.003904	1.60	0.109	.9986202	1.013924
exper	1.004782	.0049147	0.98	0.329	.9951951	1.014461
male	1.062484	.0706455	0.91	0.362	.9326637	1.210374
age	.9747149	.0041512	-6.01	0.000	.9666126	.9828851
pyacctrati~2	.9263203	.0615988	-1.15	0.250	.8131256	1.055273
pyperstudlep	1.019006	.0044942	4.27	0.000	1.010235	1.027852
pyperstude~s	.9888786	.0022991	-4.81	0.000	.9843827	.9933951
pyperstudw~e	.9971483	.0060167	-0.47	0.636	.9854253	1.009011
pyperstudb~k	1.002596	.0067982	0.38	0.702	.9893602	1.016009
pyperstudl~o	1.00214	.0064789	0.33	0.741	.9895218	1.014919
pyperstudm~e	.9995191	.0033128	-0.15	0.885	.9930472	1.006033
Asian	.5587688	.2913891	-1.12	0.264	.2010685	1.552817
Black	1.198545	.1366037	1.59	0.112	.9586033	1.498545
Latino	1.775144	.2009453	5.07	0.000	1.42193	2.216099
Other	.4411336	.1876102	-1.92	0.054	.1916733	1.015263
AsianMale	2.482318	1.996441	1.13	0.258	.5131754	12.0074
BlackMale	2.002288	.3170419	4.38	0.000	1.468072	2.730899
LatinoMale	1.254644	.1779611	1.60	0.110	.9501333	1.656749
OtherMale	2.251551	1.281409	1.43	0.154	.7379784	6.869416
3						
tenure	1.001502	.0047518	0.32	0.752	.9922317	1.010859
exper	1.060575	.0077266	8.07	0.000	1.045539	1.075828
male	1.810628	.1688706	6.37	0.000	1.508137	2.17379
age	.9661638	.0066359	-5.01	0.000	.9532448	.9792578
pyacctrati~2	1.095837	.0992135	1.01	0.312	.9176584	1.308612
pyperstudlep	1.005887	.0068952	0.86	0.392	.9924631	1.019493
pyperstude~s	1.00942	.0031053	3.05	0.002	1.003352	1.015524
pyperstudw~e	1.050725	.0122911	4.23	0.000	1.026909	1.075093
pyperstudb~k	1.038693	.0132928	2.97	0.003	1.012964	1.065076
pyperstudl~o	1.036129	.0126817	2.90	0.004	1.011569	1.061285
pyperstudm~e	.9974675	.0047316	-0.53	0.593	.9882368	1.006784
Asian	.5069805	.5276874	-0.65	0.514	.065921	3.899051
Black	.6398012	.1502346	-1.90	0.057	.403804	1.013723
Latino	1.365675	.2571514	1.66	0.098	.9442117	1.975265
Other	.4866836	.3596147	-0.97	0.330	.1143633	2.071127
AsianMale	2.26e-06	.0018265	-0.02	0.987	0	.
BlackMale	2.46584	.690347	3.22	0.001	1.424488	4.268457
LatinoMale	.9830655	.2171736	-0.08	0.938	.6375871	1.515742
OtherMale	3.167661	2.704616	1.35	0.177	.5942439	16.88545

## 2007-08

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
> Male LatinoMale OtherMale if validcert2008a==1, rrr iter(20)
```

Iteration 0: log likelihood = -26142.462  
Iteration 1: log likelihood = -24986.71  
Iteration 2: log likelihood = -24964.867  
Iteration 3: log likelihood = -24964.831  
Iteration 4: log likelihood = -24964.831

Multinomial logistic regression

Number of obs = 24470  
 LR chi2(38) = 2355.26  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0450

Log likelihood = -24964.831

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.01372	.0024406	5.66	0.000	1.008948	1.018515
	exper	1.001704	.0030258	0.56	0.573	.995791	1.007652
	male	1.174249	.0476927	3.95	0.000	1.084397	1.271546
	age	.9760448	.0023976	-9.87	0.000	.9713568	.9807554
	pyacctrati~2	.9735339	.0244177	-1.07	0.285	.9268333	1.022588
	pyperstudlep	1.005566	.001253	4.45	0.000	1.003113	1.008025
	pyperstude~s	.9946571	.0012519	-4.26	0.000	.9922064	.9971138
	pyperstudw~e	.9897615	.003286	-3.10	0.002	.9833419	.9962229
	pyperstudb~k	.9912055	.0037214	-2.35	0.019	.9839385	.9985262
	pyperstudl~o	.9921222	.0035093	-2.24	0.025	.9852679	.9990242
	pyperstudm~e	1.0044	.0021899	2.01	0.044	1.000117	1.008701
	Asian	.6655645	.1664963	-1.63	0.104	.4076197	1.086739
	Black	1.001992	.0593718	0.03	0.973	.8921288	1.125385
	Latino	1.405101	.0752543	6.35	0.000	1.265083	1.560617
	Other	1.178655	.2144603	0.90	0.366	.8251021	1.683704
	AsianMale	1.928541	.8087168	1.57	0.117	.8477839	4.387052
	BlackMale	1.68441	.159795	5.50	0.000	1.39861	2.028611
	LatinoMale	1.115677	.0877643	1.39	0.164	.9562665	1.301661
	OtherMale	.9096181	.2701957	-0.32	0.750	.5081778	1.628181
3							
	tenure	1.016311	.0023769	6.92	0.000	1.011663	1.020981
	exper	1.058673	.0035822	16.85	0.000	1.051676	1.065717
	male	1.410887	.0604754	8.03	0.000	1.2972	1.534538
	age	.9743314	.0028852	-8.78	0.000	.968693	.9800027
	pyacctrati~2	1.415425	.0373904	13.15	0.000	1.344006	1.490639
	pyperstudlep	1.010594	.0013588	7.84	0.000	1.007934	1.013261
	pyperstude~s	1.020146	.0014026	14.51	0.000	1.017401	1.022899
	pyperstudw~e	1.006092	.0039297	1.56	0.120	.9984195	1.013824
	pyperstudb~k	.9914547	.00438	-1.94	0.052	.9829071	1.000077
	pyperstudl~o	.9873556	.004102	-3.06	0.002	.9793485	.9954282
	pyperstudm~e	.9925855	.0026317	-2.81	0.005	.9874409	.9977569
	Asian	.7497449	.2166956	-1.00	0.319	.4254949	1.321091
	Black	.7869322	.0556728	-3.39	0.001	.6850428	.9039761
	Latino	1.147185	.0699067	2.25	0.024	1.018036	1.292717
	Other	.8001997	.1882633	-0.95	0.343	.5045868	1.268998
	AsianMale	1.50488	.7492452	0.82	0.412	.5671635	3.992965
	BlackMale	1.362528	.1573195	2.68	0.007	1.086586	1.708545
	LatinoMale	.9836985	.0880877	-0.18	0.854	.8253517	1.172425
	OtherMale	1.235674	.440193	0.59	0.552	.6147155	2.483896

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=24470)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	58.467	2	0.000
exper	329.425	2	0.000
male	65.756	2	0.000
age	135.443	2	0.000
pyacctrati~2	215.004	2	0.000

pyperstudlep		63.600	2	0.000
pyperstude~s		317.631	2	0.000
pyperstudw~e		18.210	2	0.000
pyperstudb~k		6.896	2	0.032
pyperstudl~o		10.750	2	0.005
pyperstudm~e		18.125	2	0.000
Asian		3.032	2	0.220
Black		13.113	2	0.001
Latino		40.282	2	0.000
Other		2.573	2	0.276
AsianMale		2.557	2	0.278
BlackMale		30.866	2	0.000
LatinoMale		2.626	2	0.269
OtherMale		0.660	2	0.719

\*\*\*\* Wald tests for independent variables (N=24470)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		58.372	2	0.000
exper		309.775	2	0.000
male		65.747	2	0.000
age		134.027	2	0.000
pyaccrati~2		215.959	2	0.000
pyperstudlep		63.240	2	0.000
pyperstude~s		311.277	2	0.000
pyperstudw~e		18.306	2	0.000
pyperstudb~k		6.857	2	0.032
pyperstudl~o		10.750	2	0.005
pyperstudm~e		17.754	2	0.000
Asian		2.980	2	0.225
Black		12.852	2	0.002
Latino		40.354	2	0.000
Other		2.522	2	0.283
AsianMale		2.528	2	0.283
BlackMale		30.615	2	0.000
LatinoMale		2.628	2	0.269
OtherMale		0.661	2	0.719

\*\*\*\* Hausman tests of IIA assumption (N=24470)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-80.674	19	---	---
3		-5.655	19	---	---

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=24470)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-5091.559	-5082.543	18.031	20	0.585	for Ho
3		-6093.465	-6084.338	18.254	20	0.571	for Ho

\*\*\*\* Wald tests for combining alternatives (N=24470)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1521.503	19	0.000
2-	1	441.397	19	0.000
3-	1	1410.823	19	0.000

\*\*\*\* LR tests for combining alternatives (N=24470)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1685.846	19	0.000
2-	1	453.005	19	0.000
3-	1	1560.011	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
> AsianMale BlackMale LatinoMale OtherMale if validcert2008a==1, rrr iter(20)
```

-----> region = 1

note: OtherMale omitted because of collinearity

```
Iteration 0: log likelihood = -1883.4244
Iteration 1: log likelihood = -1787.7794
Iteration 2: log likelihood = -1784.4505
Iteration 3: log likelihood = -1784.0971
Iteration 4: log likelihood = -1784.0424
Iteration 5: log likelihood = -1784.0291
Iteration 6: log likelihood = -1784.0263
Iteration 7: log likelihood = -1784.0256
Iteration 8: log likelihood = -1784.0255
Iteration 9: log likelihood = -1784.0255
```

Multinomial logistic regression	Number of obs	=	1750
	LR chi2(36)	=	198.80
	Prob > chi2	=	0.0000
Log likelihood = -1784.0255	Pseudo R2	=	0.0528

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1						
	tenure	.988966	.0086041	-1.28	0.202	.9722453 1.005974
	exper	1.00899	.0126209	0.72	0.474	.9845539 1.034032
	male	1.250119	.3941744	0.71	0.479	.6738439 2.319227
	age	1.018672	.0107246	1.76	0.079	.9978673 1.03991
	pyacctrati~2	1.170839	.1109727	1.66	0.096	.9723443 1.409855
	pyperstudlep	.9968319	.0029362	-1.08	0.281	.9910936 1.002603
	pyperstude~s	1.001784	.0061686	0.29	0.772	.989766 1.013947
	pyperstudw~e	.9756696	.0716602	-0.34	0.737	.8448594 1.126733
	pyperstudb~k	.8324132	.1152253	-1.33	0.185	.6346187 1.091855
	pyperstudl~o	.9391155	.0631526	-0.93	0.350	.8231488 1.07142
	pyperstudm~e	1.009676	.0105231	0.92	0.356	.9892606 1.030513
	Asian	6.040769	6.758632	1.61	0.108	.6741196 54.13118
	Black	3608806	4.49e+09	0.01	0.990	0 .
	Latino	1.055734	.2309151	0.25	0.804	.6876631 1.620814



Other		1.033792	.8239063	0.04	0.967	.2167936	4.929689
AsianMale		.1867721	1353.199	-0.00	1.000	0	.
BlackMale		1.11e-07	.0001375	-0.01	0.990	0	.
LatinoMale		.6230775	.2119396	-1.39	0.164	.3198941	1.213607
OtherMale		(omitted)					
-----							
2		(base outcome)					
-----							
3							
tenure		1.011665	.0098984	1.19	0.236	.9924496	1.031253
exper		1.098715	.0190776	5.42	0.000	1.061953	1.13675
male		.6888411	.2645343	-0.97	0.332	.3245142	1.462192
age		.9549391	.0146811	-3.00	0.003	.9265938	.9841515
pyacctrati~2		1.454874	.1532798	3.56	0.000	1.183439	1.788564
pyperstudlep		1.014886	.0034382	4.36	0.000	1.00817	1.021648
pyperstude~s		.9937953	.0072853	-0.85	0.396	.9796186	1.008177
pyperstudw~e		1.077182	.0999684	0.80	0.423	.8980343	1.292068
pyperstudb~k		1.017036	.1694365	0.10	0.919	.7337151	1.409759
pyperstudl~o		1.048182	.0866595	0.57	0.569	.8913802	1.232568
pyperstudm~e		.9927528	.0135535	-0.53	0.594	.9665407	1.019676
Asian		5.93e-07	.0007615	-0.01	0.991	0	.
Black		2786604	3.46e+09	0.01	0.990	0	.
Latino		.6315921	.1489026	-1.95	0.051	.3978868	1.002568
Other		.3431772	.4121294	-0.89	0.373	.0326055	3.611986
AsianMale		1.15e+14	6.07e+17	0.01	0.995	0	.
BlackMale		2.67e-13	4.50e-10	-0.02	0.986	0	.
LatinoMale		1.398083	.576846	0.81	0.417	.6227717	3.138606
OtherMale		(omitted)					
-----							

-----> region = 2

Iteration 0: log likelihood = -762.86823  
Iteration 1: log likelihood = -715.3235  
Iteration 2: log likelihood = -713.39001  
Iteration 3: log likelihood = -713.19516  
Iteration 4: log likelihood = -713.16282  
Iteration 5: log likelihood = -713.1555  
Iteration 6: log likelihood = -713.15369  
Iteration 7: log likelihood = -713.15334  
Iteration 8: log likelihood = -713.15328  
Iteration 9: log likelihood = -713.15327

Multinomial logistic regression	Number of obs	=	740
	LR chi2(38)	=	99.43
	Prob > chi2	=	0.0000
Log likelihood = -713.15327	Pseudo R2	=	0.0652

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.016641	.0156204	1.07	0.283	.9864816 1.047722
exper		.9813219	.0184198	-1.00	0.315	.9458757 1.018096
male		1.382925	.400512	1.12	0.263	.7839353 2.439591
age		.9816892	.0142684	-1.27	0.204	.9541182 1.010057
pyacctrati~2		.9986069	.1610307	-0.01	0.993	.7280041 1.369794
pyperstudlep		.9596725	.0215186	-1.84	0.066	.9184102 1.002789
pyperstude~s		1.01294	.0083034	1.57	0.117	.9967958 1.029346
pyperstudw~e		.8925876	.0732927	-1.38	0.166	.7599001 1.048444
pyperstudb~k		.8745449	.0785421	-1.49	0.136	.7333926 1.042864
pyperstudl~o		.892405	.0708246	-1.43	0.151	.763849 1.042597
pyperstudm~e		1.006932	.0092882	0.75	0.454	.988891 1.025302
Asian		2.792854	4.096058	0.70	0.484	.157637 49.48098

Black		1.365382	1.232524	0.35	0.730	.2327437	8.009957
Latino		1.687051	.4454128	1.98	0.048	1.005531	2.830488
Other		.9420387	1.117072	-0.05	0.960	.092196	9.625548
AsianMale		5.06e-07	.0007338	-0.01	0.992	0	.
BlackMale		1.126145	1.473113	0.09	0.928	.0867234	14.62355
LatinoMale		1.142035	.4534073	0.33	0.738	.5244912	2.486686
OtherMale		1.688298	2.555946	0.35	0.729	.0868558	32.81702
-----							
3							
tenure		1.018413	.0139507	1.33	0.183	.991434	1.046126
exper		1.027702	.0191411	1.47	0.142	.9908624	1.065911
male		.6566327	.2023183	-1.37	0.172	.358967	1.201131
age		.9835093	.015318	-1.07	0.286	.9539403	1.013995
pyacctrati~2		1.309411	.2065783	1.71	0.088	.9611401	1.783878
pyperstudlep		1.012418	.0189557	0.66	0.510	.9759392	1.050261
pyperstude~s		1.024179	.0087543	2.80	0.005	1.007164	1.041482
pyperstudw~e		1.05678	.0928405	0.63	0.530	.8896198	1.255349
pyperstudb~k		.9621121	.0973146	-0.38	0.703	.7890951	1.173065
pyperstudl~o		1.040144	.0886119	0.46	0.644	.8801931	1.229162
pyperstudm~e		.9766293	.0145046	-1.59	0.111	.9486105	1.005476
Asian		2.16e-06	.0031149	-0.01	0.993	0	.
Black		1.929277	1.520208	0.83	0.404	.4117896	9.038864
Latino		.9306214	.2393781	-0.28	0.780	.562113	1.540716
Other		.6007121	.7236748	-0.42	0.672	.0566537	6.369489
AsianMale		2030494	2.93e+09	0.01	0.992	0	.
BlackMale		.9515466	1.362454	-0.03	0.972	.057498	15.74734
LatinoMale		2.992719	1.247974	2.63	0.009	1.321637	6.776724
OtherMale		2.55e-06	.0020801	-0.02	0.987	0	.
-----							

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -320.67722  
Iteration 1: log likelihood = -277.5113  
Iteration 2: log likelihood = -276.4267  
Iteration 3: log likelihood = -276.4191  
Iteration 4: log likelihood = -276.4191

Multinomial logistic regression	Number of obs	=	297
	LR chi2(32)	=	88.52
	Prob > chi2	=	0.0000
Log likelihood = -276.4191	Pseudo R2	=	0.1380

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.075361	.0293018	2.67	0.008	1.019438 1.134353
exper		.8860752	.0275754	-3.89	0.000	.8336437 .9418044
male		3.010652	1.262205	2.63	0.009	1.323724 6.847368
age		1.027904	.0250617	1.13	0.259	.9799393 1.078217
pyacctrati~2		1.38937	.4757312	0.96	0.337	.7101658 2.718166
pyperstudlep		1.036428	.039976	0.93	0.354	.9609645 1.117817
pyperstude~s		1.003788	.020218	0.19	0.851	.9649331 1.044207
pyperstudw~e		.9522441	.0637107	-0.73	0.465	.8352144 1.085672
pyperstudb~k		.9972644	.0687683	-0.04	0.968	.8711924 1.141581
pyperstudl~o		.9573747	.0693626	-0.60	0.548	.8306378 1.103449
pyperstudm~e		1.117678	.0389281	3.19	0.001	1.043927 1.19664
Asian		(omitted)				
Black		.6085922	.5751557	-0.53	0.599	.0954752 3.87938
Latino		2.517958	1.756254	1.32	0.186	.6417156 9.87994

Other		1.619376	2.422735	0.32	0.747	.0862726	30.39645
AsianMale		(omitted)					
BlackMale		4.334224	7.137491	0.89	0.373	.1718597	109.3072
LatinoMale		.2274137	.2274339	-1.48	0.139	.0320287	1.614706
OtherMale		(omitted)					
-----							
3							
tenure		1.047058	.0218946	2.20	0.028	1.005013	1.090862
exper		1.000646	.0290678	0.02	0.982	.9452655	1.059271
male		2.679554	1.023461	2.58	0.010	1.267489	5.664748
age		.9880942	.0255063	-0.46	0.643	.9393462	1.039372
pyacctrati~2		1.41252	.3952231	1.23	0.217	.81626	2.444336
pyperstudlep		1.031292	.0342803	0.93	0.354	.9662454	1.100717
pyperstude~s		1.064642	.0194301	3.43	0.001	1.027233	1.103413
pyperstudw~e		1.009089	.0612542	0.15	0.882	.8959	1.136579
pyperstudb~k		.9827148	.0621716	-0.28	0.783	.8681126	1.112446
pyperstudl~o		.9588329	.0626239	-0.64	0.520	.8436236	1.089776
pyperstudm~e		1.04939	.033396	1.51	0.130	.9859349	1.11693
Asian		(omitted)					
Black		1.255596	.9141314	0.31	0.755	.3013952	5.230746
Latino		3.294863	2.220906	1.77	0.077	.8791926	12.34783
Other		2.826532	4.247725	0.69	0.489	.1486157	53.75803
AsianMale		(omitted)					
BlackMale		3.645579	5.097189	0.93	0.355	.2353047	56.48101
LatinoMale		.2510182	.239372	-1.45	0.147	.0387254	1.627101
OtherMale		(omitted)					
-----							

-----> region = 4

Iteration 0: log likelihood = -5405.0461  
Iteration 1: log likelihood = -5101.7733  
Iteration 2: log likelihood = -5088.6374  
Iteration 3: log likelihood = -5088.5866  
Iteration 4: log likelihood = -5088.5866

Multinomial logistic regression	Number of obs	=	5221
	LR chi2(38)	=	632.92
	Prob > chi2	=	0.0000
Log likelihood = -5088.5866	Pseudo R2	=	0.0585

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.028599	.0054377	5.33	0.000	1.017997 1.039312
exper		1.012198	.0068507	1.79	0.073	.9988592 1.025714
male		1.079106	.0981834	0.84	0.403	.9028522 1.289767
age		.9688187	.0052067	-5.89	0.000	.9586673 .9790775
pyacctrati~2		.9688035	.0501979	-0.61	0.541	.8752482 1.072359
pyperstudlep		.9972904	.00281	-0.96	0.336	.9917981 1.002813
pyperstude~s		.9988931	.0034776	-0.32	0.750	.9921003 1.005732
pyperstudw~e		.9955377	.005175	-0.86	0.390	.9854464 1.005732
pyperstudb~k		.9912615	.0063351	-1.37	0.170	.9789223 1.003756
pyperstudl~o		.9961523	.0061427	-0.63	0.532	.9841853 1.008265
pyperstudm~e		1.004799	.0046646	1.03	0.302	.9956975 1.013983
Asian		.5467185	.2296361	-1.44	0.151	.2400141 1.245348
Black		.853513	.0863418	-1.57	0.117	.7000066 1.040682
Latino		1.573632	.1925228	3.71	0.000	1.238125 2.000054
Other		1.102587	.391798	0.27	0.783	.5494691 2.212496
AsianMale		1.435002	1.007626	0.51	0.607	.3623749 5.6826
BlackMale		2.141787	.3503913	4.66	0.000	1.554253 2.951419
LatinoMale		.9919955	.2016442	-0.04	0.968	.6660164 1.477524
OtherMale		.854744	.5797028	-0.23	0.817	.2262225 3.229507

-----+-----						
3						
tenure		1.040265	.0061084	6.72	0.000	1.028362 1.052307
exper		1.076742	.0095649	8.32	0.000	1.058157 1.095653
male		1.234433	.1373054	1.89	0.058	.9926333 1.535134
age		.9597101	.0073174	-5.39	0.000	.9454749 .9741596
pyacctrati~2		1.325202	.0819993	4.55	0.000	1.17385 1.496069
pyperstudlep		1.014128	.003475	4.09	0.000	1.00734 1.020962
pyperstude~s		1.029678	.0045274	6.65	0.000	1.020843 1.03859
pyperstudw~e		1.006463	.0066805	0.97	0.332	.9934543 1.019642
pyperstudb~k		.9881955	.0081987	-1.43	0.152	.9722563 1.004396
pyperstudl~o		.9748665	.0076857	-3.23	0.001	.9599187 .9900471
pyperstudm~e		.9903283	.0062343	-1.54	0.123	.9781844 1.002623
Asian		.8225969	.3843156	-0.42	0.676	.3292361 2.05526
Black		.5245498	.0707957	-4.78	0.000	.4026286 .6833904
Latino		1.227317	.1862689	1.35	0.177	.9115296 1.652506
Other		.59139	.335607	-0.93	0.355	.1944585 1.798544
AsianMale		.8854669	.8233619	-0.13	0.896	.1431108 5.478633
BlackMale		1.820177	.4006156	2.72	0.007	1.182406 2.801951
LatinoMale		1.01885	.2646545	0.07	0.943	.6123556 1.695183
OtherMale		2.179066	1.952193	0.87	0.385	.3764363 12.6139
-----+-----						

-----> region = 5

Iteration 0: log likelihood = -533.02433  
Iteration 1: log likelihood = -489.00507  
Iteration 2: log likelihood = -487.39945  
Iteration 3: log likelihood = -487.23444  
Iteration 4: log likelihood = -487.19665  
Iteration 5: log likelihood = -487.18916  
Iteration 6: log likelihood = -487.18782  
Iteration 7: log likelihood = -487.18752  
Iteration 8: log likelihood = -487.18745  
Iteration 9: log likelihood = -487.18743

Multinomial logistic regression	Number of obs	=	512
	LR chi2(38)	=	91.67
	Prob > chi2	=	0.0000
Log likelihood = -487.18743	Pseudo R2	=	0.0860

-----+-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
1		(base outcome)				
-----+-----						
2						
tenure		1.017714	.0189683	0.94	0.346	.981208 1.055579
exper		.9517726	.0230565	-2.04	0.041	.9076387 .9980525
male		1.934618	.5452606	2.34	0.019	1.113497 3.361254
age		1.007242	.0186719	0.39	0.697	.9713022 1.044511
pyacctrati~2		.8951155	.1778848	-0.56	0.577	.606347 1.321408
pyperstudlep		.9864265	.0335009	-0.40	0.687	.9229036 1.054322
pyperstude~s		.9979882	.0107833	-0.19	0.852	.9770756 1.019348
pyperstudw~e		1.08711	.0581807	1.56	0.119	.9788544 1.207337
pyperstudb~k		1.081266	.0586719	1.44	0.150	.9721753 1.202599
pyperstudl~o		1.088685	.0712468	1.30	0.194	.9576283 1.237677
pyperstudm~e		.9895009	.0241659	-0.43	0.666	.9432523 1.038017
Asian		6.101005	9.731735	1.13	0.257	.2676989 139.0452
Black		1.79887	.7443613	1.42	0.156	.7994252 4.047825
Latino		1.361304	1.642035	0.26	0.798	.1280021 14.47748
Other		10.5387	16.2344	1.53	0.126	.5146946 215.7868
AsianMale		.133497	.2877193	-0.93	0.350	.001954 9.12059
BlackMale		.9961422	.53672	-0.01	0.994	.3464935 2.863832
LatinoMale		.9870176	3746.589	-0.00	1.000	0 .
OtherMale		561112.6	1.24e+09	0.01	0.995	0 .

-----+-----						
3						
tenure		1.002744	.0156247	0.18	0.860	.9725834 1.033841
exper		1.078981	.0287982	2.85	0.004	1.023989 1.136927
male		3.137191	.8988577	3.99	0.000	1.789194 5.500782
age		.9583814	.0225457	-1.81	0.071	.915196 1.003605
pyacctrati~2		1.308799	.2488863	1.42	0.157	.9015816 1.899943
pyperstudlep		1.010699	.033507	0.32	0.748	.9471143 1.078552
pyperstude~s		1.022334	.0107834	2.09	0.036	1.001416 1.043689
pyperstudw~e		1.02556	.0466711	0.55	0.579	.9380471 1.121237
pyperstudb~k		1.00137	.0461304	0.03	0.976	.9149179 1.095992
pyperstudl~o		1.00541	.0589078	0.09	0.927	.8963357 1.127758
pyperstudm~e		1.003043	.0235187	0.13	0.897	.9579905 1.050215
Asian		4.92e-06	.0051921	-0.01	0.991	0 .
Black		2.554061	1.061027	2.26	0.024	1.131406 5.76559
Latino		2.161602	2.63063	0.63	0.526	.1990118 23.47861
Other		3.22e-06	.0033749	-0.01	0.990	0 .
AsianMale		.218419	329.1312	-0.00	0.999	0 .
BlackMale		.4794287	.2785786	-1.27	0.206	.1535048 1.49736
LatinoMale		2032773	4.46e+09	0.01	0.995	0 .
OtherMale		195613	7.76e+08	0.00	0.998	0 .
-----+-----						

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -956.75684  
Iteration 1: log likelihood = -900.89447  
Iteration 2: log likelihood = -899.58392  
Iteration 3: log likelihood = -899.41205  
Iteration 4: log likelihood = -899.38288  
Iteration 5: log likelihood = -899.38003  
Iteration 6: log likelihood = -899.37943  
Iteration 7: log likelihood = -899.37931  
Iteration 8: log likelihood = -899.37928  
Iteration 9: log likelihood = -899.37927

Multinomial logistic regression	Number of obs	=	893
	LR chi2(36)	=	114.76
	Prob > chi2	=	0.0000
Log likelihood = -899.37927	Pseudo R2	=	0.0600

-----+-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
1		(base outcome)				
-----+-----						
2						
tenure		1.002326	.0131608	0.18	0.860	.9768607 1.028456
exper		1.019567	.0157995	1.25	0.211	.9890665 1.051009
male		1.240392	.2282431	1.17	0.242	.864829 1.779047
age		.965222	.012257	-2.79	0.005	.9414952 .9895467
pyacctrati~2		.8326145	.1013204	-1.51	0.132	.6559363 1.056881
pyperstudlep		1.027363	.014373	1.93	0.054	.9995751 1.055923
pyperstude~s		.9927404	.0065169	-1.11	0.267	.9800492 1.005596
pyperstudw~e		1.026019	.0289448	0.91	0.363	.9708284 1.084348
pyperstudb~k		1.011479	.0310796	0.37	0.710	.9523624 1.074266
pyperstudl~o		1.018169	.0293253	0.63	0.532	.9622847 1.077299
pyperstudm~e		1.003923	.0133513	0.29	0.768	.9780933 1.030436
Asian		1.526216	2.202375	0.29	0.770	.090221 25.81813
Black		2.641575	1.064187	2.41	0.016	1.199354 5.818065
Latino		.4310029	.2936884	-1.24	0.217	.1133615 1.638683
Other		3.18e-07	.00045	-0.01	0.992	0 .
AsianMale		(omitted)				
BlackMale		.8029029	.5245529	-0.34	0.737	.2231288 2.889152
LatinoMale		2.831158	2.49219	1.18	0.237	.5042838 15.89473

OtherMale		4.66e+13	1.41e+17	0.01	0.992	0	.
-----							
3							
tenure		1.009952	.0132647	0.75	0.451	.9842852	1.036288
exper		1.066185	.0187736	3.64	0.000	1.030017	1.103623
male		2.711261	.5378625	5.03	0.000	1.837846	3.999756
age		.9695052	.0146991	-2.04	0.041	.9411193	.9987472
pyacctrati~2		1.41078	.1866758	2.60	0.009	1.088498	1.828484
pyperstudlep		1.027827	.0160174	1.76	0.078	.9969086	1.059705
pyperstude~s		1.01795	.0070578	2.57	0.010	1.004211	1.031878
pyperstudw~e		1.001616	.0276592	0.06	0.953	.948846	1.057321
pyperstudb~k		.9954095	.0296654	-0.15	0.877	.9389319	1.055284
pyperstudl~o		.9746843	.0275023	-0.91	0.363	.9222441	1.030106
pyperstudm~e		1.012981	.0131486	0.99	0.320	.9875357	1.039083
Asian		1.27e-06	.0018516	-0.01	0.993	0	.
Black		1.746479	.8375985	1.16	0.245	.6822342	4.470884
Latino		.8877678	.6144395	-0.17	0.863	.2286483	3.446916
Other		3.03e-07	.0004743	-0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		.414463	.3268346	-1.12	0.264	.088359	1.94411
LatinoMale		.9150041	.8759849	-0.09	0.926	.1401264	5.974838
OtherMale		829166.5	3.89e+09	0.00	0.998	0	.
-----							

-----> region = 7

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1123.7493  
Iteration 1: log likelihood = -1056.406  
Iteration 2: log likelihood = -1054.4347  
Iteration 3: log likelihood = -1054.3646  
Iteration 4: log likelihood = -1054.3487  
Iteration 5: log likelihood = -1054.3451  
Iteration 6: log likelihood = -1054.3444  
Iteration 7: log likelihood = -1054.3443  
Iteration 8: log likelihood = -1054.3442

Multinomial logistic regression	Number of obs	=	1078
	LR chi2(34)	=	138.81
	Prob > chi2	=	0.0000
Log likelihood = -1054.3442	Pseudo R2	=	0.0618

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9948321	.0121681	-0.42	0.672	.9712667 1.018969
exper		.9907974	.0148403	-0.62	0.537	.9621337 1.020315
male		1.262749	.2371734	1.24	0.214	.8738595 1.824703
age		.9817134	.0125189	-1.45	0.148	.9574809 1.006559
pyacctrati~2		1.025413	.1749103	0.15	0.883	.7340149 1.432495
pyperstudlep		1.012516	.0144577	0.87	0.384	.9845726 1.041253
pyperstude~s		.9851516	.0089671	-1.64	0.100	.9677323 1.002884
pyperstudw~e		.9038524	.0701525	-1.30	0.193	.7763035 1.052358
pyperstudb~k		.9124498	.0723246	-1.16	0.248	.7811585 1.065808
pyperstudl~o		.9207176	.0727689	-1.05	0.296	.7885909 1.074982
pyperstudm~e		.9609154	.0177586	-2.16	0.031	.9267321 .9963596
Asian		(omitted)				
Black		1.299169	.4201203	0.81	0.418	.6892983 2.448634
Latino		.6663091	.5780771	-0.47	0.640	.1216717 3.648901
Other		2.830136	3.544398	0.83	0.406	.2430983 32.94828
AsianMale		(omitted)				
BlackMale		1.729205	.8171149	1.16	0.246	.6848907 4.365876

LatinoMale		3.588589	3.749468	1.22	0.221	.4629802	27.81539
OtherMale		4.81e-07	.0004638	-0.02	0.988	0	.
-----							
3							
tenure		.9917175	.010015	-0.82	0.410	.9722814	1.011542
exper		1.033163	.0140176	2.40	0.016	1.006052	1.061006
male		1.588586	.272454	2.70	0.007	1.135073	2.2233
age		.9882374	.012002	-0.97	0.330	.9649917	1.012043
pyacctrati~2		1.518105	.2159756	2.93	0.003	1.148694	2.006316
pyperstudlep		1.039183	.0151817	2.63	0.009	1.009849	1.069368
pyperstude~s		1.034875	.0078965	4.49	0.000	1.019513	1.050468
pyperstudw~e		.8626745	.0660558	-1.93	0.054	.7424542	1.002361
pyperstudb~k		.8522224	.0664844	-2.05	0.040	.7313886	.9930194
pyperstudl~o		.8259506	.0642162	-2.46	0.014	.7092098	.9619079
pyperstudm~e		.9911042	.013234	-0.67	0.503	.9655026	1.017385
Asian		(omitted)					
Black		1.337275	.4175478	0.93	0.352	.725176	2.466028
Latino		.6818566	.4943384	-0.53	0.597	.1646575	2.823609
Other		2.368134	3.42771	0.60	0.551	.1387876	40.40748
AsianMale		(omitted)					
BlackMale		.980682	.4679851	-0.04	0.967	.3848856	2.498761
LatinoMale		1.723927	1.742223	0.54	0.590	.2378392	12.49552
OtherMale		6.09e-07	.0005126	-0.02	0.986	0	.
-----							

-----> region = 8

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -397.35957  
Iteration 1: log likelihood = -358.28999  
Iteration 2: log likelihood = -356.80258  
Iteration 3: log likelihood = -356.69057  
Iteration 4: log likelihood = -356.66601  
Iteration 5: log likelihood = -356.66082  
Iteration 6: log likelihood = -356.65988  
Iteration 7: log likelihood = -356.65979  
Iteration 8: log likelihood = -356.65977

Multinomial logistic regression	Number of obs	=	376
	LR chi2(34)	=	81.40
	Prob > chi2	=	0.0000
Log likelihood = -356.65977	Pseudo R2	=	0.1024

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
1	(base outcome)					
-----						
2						
tenure	1.05466	.0241001	2.33	0.020	1.008467	1.10297
exper	1.007406	.030282	0.25	0.806	.9497684	1.06854
male	3.358642	1.114235	3.65	0.000	1.752977	6.435041
age	.9370445	.0251432	-2.42	0.015	.8890381	.9876431
pyacctrati~2	1.034837	.2310392	0.15	0.878	.6680843	1.602922
pyperstudlep	1.023916	.032255	0.75	0.453	.9626096	1.089127
pyperstude~s	.9698117	.0136108	-2.18	0.029	.9434986	.9968586
pyperstudw~e	1.016316	.1533125	0.11	0.915	.7561778	1.365946
pyperstudb~k	1.054486	.1597057	0.35	0.726	.7836519	1.418921
pyperstudl~o	1.032145	.1565492	0.21	0.835	.7667182	1.389458
pyperstudm~e	1.05753	.0385219	1.54	0.125	.984661	1.135792
Asian	.8943782	2445.191	-0.00	1.000	0	.
Black	2.362894	1.26023	1.61	0.107	.8307361	6.720871
Latino	1904972	1.16e+09	0.02	0.981	0	.
Other	3.553109	3.642495	1.24	0.216	.4764234	26.49866
AsianMale	(omitted)					

BlackMale		.3766444	.4159933	-0.88	0.377	.0432318	3.281407
LatinoMale		(omitted)					
OtherMale		1.20e-07	.0001859	-0.01	0.992	0	.

---

3							
tenure		1.005956	.0181552	0.33	0.742	.9709941	1.042176
exper		1.009069	.0222207	0.41	0.682	.9664441	1.053575
male		2.202005	.6268194	2.77	0.006	1.260423	3.846981
age		1.012119	.0206115	0.59	0.554	.972517	1.053334
pyacctrati~2		1.561151	.3045568	2.28	0.022	1.065095	2.288239
pyperstudlep		.9871415	.0323761	-0.39	0.693	.925682	1.052681
pyperstude~s		1.027709	.0125016	2.25	0.025	1.003496	1.052506
pyperstudw~e		1.022583	.1321216	0.17	0.863	.7938165	1.317277
pyperstudb~k		1.004219	.1301372	0.03	0.974	.7789702	1.294601
pyperstudl~o		1.00822	.1315648	0.06	0.950	.780693	1.302058
pyperstudm~e		1.017743	.0304743	0.59	0.557	.9597335	1.079259
Asian		1262547	1.99e+09	0.01	0.993	0	.
Black		1.697152	.9416537	0.95	0.340	.572054	5.035056
Latino		1324768	8.07e+08	0.02	0.982	0	.
Other		1.992036	2.152899	0.64	0.524	.2395292	16.5667
AsianMale		(omitted)					
BlackMale		1.146649	1.264198	0.12	0.901	.1321199	9.951604
LatinoMale		(omitted)					
OtherMale		5.34e-07	.0007123	-0.01	0.991	0	.

---

-----> region = 9

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -230.21821  
Iteration 1: log likelihood = -201.55504  
Iteration 2: log likelihood = -200.01786  
Iteration 3: log likelihood = -199.67397  
Iteration 4: log likelihood = -199.60954  
Iteration 5: log likelihood = -199.59495  
Iteration 6: log likelihood = -199.59176  
Iteration 7: log likelihood = -199.59097  
Iteration 8: log likelihood = -199.59081  
Iteration 9: log likelihood = -199.59078  
Iteration 10: log likelihood = -199.59077

Multinomial logistic regression	Number of obs	=	216
	LR chi2(34)	=	61.25
	Prob > chi2	=	0.0028
Log likelihood = -199.59077	Pseudo R2	=	0.1330

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
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1						
tenure		.9900113	.0237753	-0.42	0.676	.9444923 1.037724
exper		.9578024	.0328765	-1.26	0.209	.8954853 1.024456
male		.6691789	.2492174	-1.08	0.281	.3225043 1.38851
age		1.017086	.0302322	0.57	0.569	.9595245 1.0781
pyacctrati~2		.3707673	.113494	-3.24	0.001	.2034903 .6755526
pyperstudlep		.9962543	.0526697	-0.07	0.943	.898192 1.105023
pyperstude~s		.9581204	.0161747	-2.53	0.011	.9269374 .9903525
pyperstudw~e		1.107394	.135248	0.84	0.404	.8716528 1.406893
pyperstudb~k		1.107019	.1457796	0.77	0.440	.855191 1.433002
pyperstudl~o		1.157375	.1419522	1.19	0.233	.9100681 1.471885
pyperstudm~e		1.083604	.0433657	2.01	0.045	1.001857 1.172022
Asian		5.84e-08	.0002243	-0.00	0.997	0 .
Black		3.05e-07	.0004623	-0.01	0.992	0 .
Latino		3.53e-08	.0000963	-0.01	0.995	0 .
Other		1.203137	14645.66	0.00	1.000	0 .



AsianMale		(omitted)					
BlackMale		1.10e+13	2.46e+16	0.01	0.989	0	.
LatinoMale		(omitted)					
OtherMale		1.43e-07	.0017735	-0.00	0.999	0	.
-----							
2							
tenure		.9968959	.031613	-0.10	0.922	.9368218	1.060822
exper		.9578258	.0407586	-1.01	0.311	.881181	1.041137
male		.4240845	.1919803	-1.89	0.058	.1746301	1.029878
age		1.002081	.0340428	0.06	0.951	.937531	1.071075
pyacctrati~2		.7423427	.250795	-0.88	0.378	.3828532	1.439384
pyperstudlep		1.05025	.0638041	0.81	0.420	.9323543	1.183053
pyperstude~s		.9623777	.0195957	-1.88	0.060	.9247272	1.001561
pyperstudw~e		.8525543	.1202451	-1.13	0.258	.6466476	1.124026
pyperstudb~k		.8768588	.1328177	-0.87	0.386	.6516258	1.179943
pyperstudl~o		.8721026	.1247369	-0.96	0.339	.658901	1.15429
pyperstudm~e		1.029571	.0484238	0.62	0.536	.9389052	1.128992
Asian		9.85e-08	.0005003	-0.00	0.997	0	.
Black		.4850794	.6996586	-0.50	0.616	.0287127	8.195039
Latino		5.64e-08	.0001904	-0.00	0.996	0	.
Other		1.13e+08	9.57e+11	0.00	0.998	0	.
AsianMale		(omitted)					
BlackMale		3.23e+07	5.27e+10	0.01	0.992	0	.
LatinoMale		(omitted)					
OtherMale		2.45e-08	.0002076	-0.00	0.998	0	.
-----							
3		(base outcome)					
-----							

-----> region = 10

Iteration 0: log likelihood = -3905.3704  
Iteration 1: log likelihood = -3667.3031  
Iteration 2: log likelihood = -3661.4797  
Iteration 3: log likelihood = -3661.4568  
Iteration 4: log likelihood = -3661.4568

Multinomial logistic regression	Number of obs	=	3671
	LR chi2(38)	=	487.83
	Prob > chi2	=	0.0000
Log likelihood = -3661.4568	Pseudo R2	=	0.0625

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.017742	.0064674	2.77	0.006	1.005145 1.030497
exper		1.024293	.0082161	2.99	0.003	1.008316 1.040524
male		1.455716	.1536646	3.56	0.000	1.183653 1.790313
age		.9602176	.0062805	-6.21	0.000	.9479867 .9726064
pyacctrati~2		1.095563	.0697264	1.43	0.152	.967082 1.241114
pyperstudlep		.9994503	.0044389	-0.12	0.901	.9907879 1.008188
pyperstude~s		.9898728	.0036405	-2.77	0.006	.9827633 .9970338
pyperstudw~e		.9885639	.0073799	-1.54	0.123	.9742049 1.003135
pyperstudb~k		.9919213	.0081973	-0.98	0.326	.9759844 1.008118
pyperstudl~o		.9961496	.0080626	-0.48	0.634	.9804718 1.012078
pyperstudm~e		1.015798	.006475	2.46	0.014	1.003186 1.028568
Asian		1.136225	.6799841	0.21	0.831	.3516047 3.671755
Black		1.013956	.1307576	0.11	0.914	.787499 1.305534
Latino		1.777679	.3102161	3.30	0.001	1.262739 2.502611
Other		1.128511	.4926994	0.28	0.782	.4795982 2.655425
AsianMale		1.099094	1.0358	0.10	0.920	.1733214 6.96976
BlackMale		1.351015	.2754325	1.48	0.140	.9059931 2.01463
LatinoMale		1.146353	.3130202	0.50	0.617	.6712611 1.957696

OtherMale		.6193285	.4207403	-0.71	0.481	.1635528	2.345223
-----							
3							
tenure		1.025668	.0065325	3.98	0.000	1.012944	1.038551
exper		1.073196	.009924	7.64	0.000	1.053921	1.092824
male		1.469501	.1734916	3.26	0.001	1.165939	1.852099
age		.9639751	.0077205	-4.58	0.000	.9489615	.9792263
pyacctrati~2		1.723122	.1186251	7.90	0.000	1.505625	1.972039
pyperstudlep		1.022704	.0052799	4.35	0.000	1.012407	1.033104
pyperstude~s		1.010527	.0040301	2.63	0.009	1.002659	1.018457
pyperstudw~e		1.006393	.0081622	0.79	0.432	.9905218	1.022518
pyperstudb~k		.9968751	.0091934	-0.34	0.734	.9790183	1.015058
pyperstudl~o		.9833964	.0086273	-1.91	0.056	.9666317	1.000452
pyperstudm~e		1.006004	.0077077	0.78	0.435	.9910104	1.021225
Asian		1.267748	.9002705	0.33	0.738	.3151852	5.099174
Black		.875384	.1374038	-0.85	0.396	.6435623	1.190712
Latino		1.906349	.3677018	3.34	0.001	1.306232	2.782175
Other		.8497981	.4658146	-0.30	0.767	.290223	2.488282
AsianMale		.5081087	.6835885	-0.50	0.615	.0363735	7.097869
BlackMale		1.541848	.3843456	1.74	0.082	.9459267	2.513192
LatinoMale		1.22067	.3755559	0.65	0.517	.6679043	2.23091
OtherMale		1.274964	.9884536	0.31	0.754	.2789834	5.826627
-----							

-----> region = 11

Iteration 0: log likelihood = -2903.2908  
Iteration 1: log likelihood = -2736.515  
Iteration 2: log likelihood = -2732.3081  
Iteration 3: log likelihood = -2732.2962  
Iteration 4: log likelihood = -2732.2962

Multinomial logistic regression	Number of obs	=	2702
	LR chi2(38)	=	341.99
	Prob > chi2	=	0.0000
Log likelihood = -2732.2962	Pseudo R2	=	0.0589

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----						
1	(base outcome)					
-----						
2						
tenure	1.007283	.0076453	0.96	0.339	.9924095	1.02238
exper	1.023876	.0093597	2.58	0.010	1.005694	1.042386
male	1.182698	.126811	1.56	0.118	.9585316	1.459289
age	.9621373	.0071323	-5.21	0.000	.9482592	.9762184
pyacctrati~2	1.00394	.0772746	0.05	0.959	.8633554	1.167417
pyperstudlep	1.001787	.0056365	0.32	0.751	.9908002	1.012895
pyperstude~s	.999011	.0043514	-0.23	0.820	.9905187	1.007576
pyperstudw~e	.9947905	.0103437	-0.50	0.615	.9747224	1.015272
pyperstudb~k	.9911408	.0118795	-0.74	0.458	.9681289	1.0147
pyperstudl~o	.9996432	.0113907	-0.03	0.975	.9775654	1.02222
pyperstudm~e	.9999619	.0075605	-0.01	0.996	.9852528	1.01489
Asian	.5033147	.4269482	-0.81	0.418	.0954528	2.653938
Black	1.129464	.2076043	0.66	0.508	.7877979	1.619309
Latino	1.407561	.3250855	1.48	0.139	.8951103	2.213389
Other	1.111403	.5211818	0.23	0.822	.4433114	2.786342
AsianMale	5.664748	8.144207	1.21	0.228	.3383827	94.83157
BlackMale	1.421497	.4317084	1.16	0.247	.7838554	2.577838
LatinoMale	1.147227	.405821	0.39	0.698	.5735151	2.294849
OtherMale	1.247838	.9423273	0.29	0.769	.2840261	5.482243
-----						
3						
tenure	1.012264	.0075663	1.63	0.103	.9975426	1.027203
exper	1.091567	.0115041	8.31	0.000	1.069251	1.114349

male		1.487223	.1811975	3.26	0.001	1.171302	1.888353
age		.9676553	.0088555	-3.59	0.000	.9504535	.9851684
pyacctrati~2		1.568893	.1351308	5.23	0.000	1.32519	1.857413
pyperstudlep		1.020335	.0068926	2.98	0.003	1.006915	1.033934
pyperstude~s		1.023682	.0050075	4.78	0.000	1.013915	1.033544
pyperstudw~e		1.014008	.0125076	1.13	0.259	.9897874	1.038821
pyperstudb~k		1.001117	.0142536	0.08	0.937	.9735672	1.029447
pyperstudl~o		.9891702	.013391	-0.80	0.421	.9632695	1.015767
pyperstudm~e		.9917566	.0089003	-0.92	0.356	.9744648	1.009355
Asian		.7119293	.6295615	-0.38	0.701	.1258104	4.028629
Black		1.174111	.2680331	0.70	0.482	.7505711	1.836651
Latino		1.716271	.4559884	2.03	0.042	1.01961	2.888935
Other		.8520146	.5290338	-0.26	0.796	.2523005	2.877239
AsianMale		5.668438	8.764127	1.12	0.262	.2737773	117.3625
BlackMale		1.404914	.5343918	0.89	0.371	.6666154	2.960902
LatinoMale		.902964	.3770911	-0.24	0.807	.3982869	2.047127
OtherMale		1.405535	1.339554	0.36	0.721	.2170696	9.100907

-----> region = 12

Iteration 0: log likelihood = -943.95923  
Iteration 1: log likelihood = -878.13905  
Iteration 2: log likelihood = -876.55561  
Iteration 3: log likelihood = -876.24667  
Iteration 4: log likelihood = -876.18918  
Iteration 5: log likelihood = -876.1761  
Iteration 6: log likelihood = -876.17322  
Iteration 7: log likelihood = -876.17251  
Iteration 8: log likelihood = -876.17237  
Iteration 9: log likelihood = -876.17233  
Iteration 10: log likelihood = -876.17233

Multinomial logistic regression	Number of obs	=	869
	LR chi2(38)	=	135.57
	Prob > chi2	=	0.0000
Log likelihood = -876.17233	Pseudo R2	=	0.0718

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.002697	.0127976	0.21	0.833	.9779255 1.028096
exper		1.021841	.016865	1.31	0.190	.9893153 1.055436
male		1.079536	.215633	0.38	0.702	.7298156 1.59684
age		.9658685	.0130977	-2.56	0.010	.9405357 .9918836
pyacctrati~2		1.107427	.1799655	0.63	0.530	.8053565 1.522797
pyperstudlep		1.013636	.0173249	0.79	0.428	.9802424 1.048167
pyperstude~s		1.004323	.008065	0.54	0.591	.9886392 1.020255
pyperstudw~e		.9374823	.0542816	-1.11	0.265	.8369072 1.050144
pyperstudb~k		.9626581	.0593016	-0.62	0.537	.8531716 1.086195
pyperstudl~o		.9460595	.0549801	-0.95	0.340	.8442111 1.060195
pyperstudm~e		.9929293	.0104259	-0.68	0.499	.9727037 1.013575
Asian		7.81e-08	.0001221	-0.01	0.992	0 .
Black		.9140984	.2902047	-0.28	0.777	.4906327 1.703058
Latino		.4481155	.2111799	-1.70	0.089	.1779306 1.128572
Other		6415401	9.74e+09	0.01	0.992	0 .
AsianMale		4.91e+14	1.64e+18	0.01	0.992	0 .
BlackMale		1.186977	.6393571	0.32	0.750	.4129985 3.411428
LatinoMale		1.957321	1.393673	0.94	0.346	.4848201 7.902122
OtherMale		3.69e-14	9.95e-11	-0.01	0.991	0 .
3						
tenure		1.000862	.0117702	0.07	0.942	.9780571 1.0242

exper		1.037038	.0166334	2.27	0.023	1.004944	1.070157
male		1.208584	.2367534	0.97	0.333	.8232509	1.774277
age		.9919227	.0137326	-0.59	0.558	.9653692	1.019207
pyacctrati~2		1.364121	.2142958	1.98	0.048	1.002615	1.855974
pyperstudlep		1.031465	.0182503	1.75	0.080	.9963086	1.067863
pyperstude~s		1.023944	.0083534	2.90	0.004	1.007702	1.040448
pyperstudw~e		1.052727	.067129	0.81	0.420	.929046	1.192873
pyperstudb~k		1.056649	.0712997	0.82	0.414	.9257502	1.206056
pyperstudl~o		1.030689	.0663248	0.47	0.639	.908558	1.169236
pyperstudm~e		.969298	.0123391	-2.45	0.014	.9454131	.9937864
Asian		.4067924	.4676292	-0.78	0.434	.0427437	3.871449
Black		.5937275	.2343713	-1.32	0.187	.2738919	1.287049
Latino		.2775836	.1831885	-1.94	0.052	.0761469	1.011896
Other		.6191391	9.40e+09	0.01	0.992	0	.
AsianMale		4.304587	19814.29	0.00	1.000	0	.
BlackMale		2.082027	1.251396	1.22	0.222	.641024	6.762362
LatinoMale		4.650813	3.946363	1.81	0.070	.8815707	24.53582
OtherMale		4.24e-07	.0006448	-0.01	0.992	0	.

-----> region = 13

```
Iteration 0: log likelihood = -1643.0697
Iteration 1: log likelihood = -1560.5627
Iteration 2: log likelihood = -1559.0799
Iteration 3: log likelihood = -1559.0215
Iteration 4: log likelihood = -1559.0143
Iteration 5: log likelihood = -1559.0134
Iteration 6: log likelihood = -1559.0132
Iteration 7: log likelihood = -1559.0132
Iteration 8: log likelihood = -1559.0131
```

Multinomial logistic regression	Number of obs	=	1507
	LR chi2(38)	=	168.11
	Prob > chi2	=	0.0000
Log likelihood = -1559.0131	Pseudo R2	=	0.0512

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9924514	.0105114	-0.72	0.474	.9720619 1.013269
exper		.968535	.0106291	-2.91	0.004	.9479248 .9895934
male		1.093236	.1712369	0.57	0.569	.8042438 1.486073
age		.9969587	.0091105	-0.33	0.739	.9792614 1.014976
pyacctrati~2		.9260815	.1012893	-0.70	0.483	.7473943 1.147489
pyperstudlep		.993765	.0071673	-0.87	0.386	.9798162 1.007912
pyperstude~s		1.000348	.0067895	0.05	0.959	.9871287 1.013744
pyperstudw~e		.9742233	.0167029	-1.52	0.128	.9420301 1.007517
pyperstudb~k		.9713583	.0211226	-1.34	0.181	.9308286 1.013653
pyperstudl~o		.9770906	.0178614	-1.27	0.205	.9427027 1.012733
pyperstudm~e		1.011719	.0104762	1.13	0.261	.9913933 1.032462
Asian		1.076516	1.101466	0.07	0.943	.1449077 7.997416
Black		1.670772	.4320645	1.98	0.047	1.006457 2.773571
Latino		1.565546	.3441564	2.04	0.041	1.017525 2.408723
Other		1.42692	1.107633	0.46	0.647	.3116469 6.533358
AsianMale		.7710957	1.35524	-0.15	0.882	.0246081 24.16232
BlackMale		.6266224	.2919438	-1.00	0.316	.2514378 1.561641
LatinoMale		1.570977	.5977674	1.19	0.235	.7452158 3.311752
OtherMale		.6111689	.634464	-0.47	0.635	.0798943 4.675271
3						
tenure		1.021115	.0099357	2.15	0.032	1.001826 1.040776
exper		1.023499	.0122215	1.95	0.052	.9998239 1.047735

male		.9782403	.1625683	-0.13	0.895	.7062998	1.354884
age		.9949673	.0105403	-0.48	0.634	.9745217	1.015842
pyacctrati~2		1.347635	.1537062	2.62	0.009	1.077674	1.685222
pyperstudlep		.9937204	.0073929	-0.85	0.397	.9793357	1.008316
pyperstude~s		1.021904	.0073252	3.02	0.003	1.007647	1.036362
pyperstudw~e		.9776634	.0182197	-1.21	0.225	.9425977	1.014034
pyperstudb~k		.9593826	.0227967	-1.75	0.081	.9157263	1.00512
pyperstudl~o		.9686381	.0191765	-1.61	0.108	.9317726	1.006962
pyperstudm~e		1.002012	.0124534	0.16	0.872	.9778989	1.02672
Asian		3.149141	2.783311	1.30	0.194	.5570231	17.80373
Black		.7737772	.2478824	-0.80	0.423	.4129813	1.449778
Latino		1.001801	.2419573	0.01	0.994	.6240183	1.608295
Other		1.657346	1.375508	0.61	0.543	.325808	8.430723
AsianMale		1.24e-06	.000965	-0.02	0.986	0	.
BlackMale		.7201359	.4478258	-0.53	0.598	.212855	2.43638
LatinoMale		2.547516	1.053386	2.26	0.024	1.132789	5.729077
OtherMale		.7081234	.7914221	-0.31	0.757	.0792096	6.330527

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -244.69089  
Iteration 1: log likelihood = -218.89126  
Iteration 2: log likelihood = -218.08391  
Iteration 3: log likelihood = -215.0936  
Iteration 4: log likelihood = -214.87739  
Iteration 5: log likelihood = -214.84864  
Iteration 6: log likelihood = -214.84205  
Iteration 7: log likelihood = -214.84072  
Iteration 8: log likelihood = -214.84048  
Iteration 9: log likelihood = -214.84043  
Iteration 10: log likelihood = -214.84041

Multinomial logistic regression	Number of obs	=	247
	LR chi2(34)	=	59.70
	Prob > chi2	=	0.0042
Log likelihood = -214.84041	Pseudo R2	=	0.1220

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.015353	.0300065	0.52	0.606	.9582122 1.075901
exper		.9876357	.0402436	-0.31	0.760	.9118272 1.069747
male		.8753482	.398238	-0.29	0.770	.3588627 2.135174
age		1.013492	.0375806	0.36	0.718	.9424479 1.089891
pyacctrati~2		.7431072	.2988089	-0.74	0.460	.3378914 1.634278
pyperstudlep		.9866157	.0775081	-0.17	0.864	.8458201 1.150848
pyperstude~s		1.008508	.0193603	0.44	0.659	.9712675 1.047176
pyperstudw~e		.8146871	.1478091	-1.13	0.259	.5708959 1.162585
pyperstudb~k		.8437778	.1725018	-0.83	0.406	.5652083 1.259643
pyperstudl~o		.8161328	.1468743	-1.13	0.259	.5735554 1.161305
pyperstudm~e		.9020772	.0546803	-1.70	0.089	.8010272 1.015875
Asian		(omitted)				
Black		.4078157	1597.581	-0.00	1.000	0 .
Latino		1.137701	2903.644	0.00	1.000	0 .
Other		7.33e-08	.0002287	-0.01	0.996	0 .
AsianMale		(omitted)				
BlackMale		1.699335	9414.41	0.00	1.000	0 .
LatinoMale		19.37951	49460.49	0.00	0.999	0 .
OtherMale		2.88e+14	1.02e+18	0.01	0.993	0 .

```

3      |
      tenure |      1.00091      .0202203      0.05      0.964      .9620535      1.041336
      exper |      1.033338      .0318276      1.06      0.287      .972803      1.097641
      male |      2.148536      .7086613      2.32      0.020      1.125607      4.10108
      age |      .975315      .0270675      -0.90      0.368      .9236808      1.029836
pyacctrati~2 |      1.628739      .4226436      1.88      0.060      .9794269      2.708513
pyperstudlep |      1.019551      .0440165      0.45      0.654      .9368293      1.109577
pyperstude~s |      1.049508      .0148418      3.42      0.001      1.020818      1.079005
pyperstudw~e |      .7843261      .1115411      -1.71      0.088      .5935333      1.03645
pyperstudb~k |      .7349939      .118205      -1.91      0.056      .5362776      1.007344
pyperstudl~o |      .7791427      .1103405      -1.76      0.078      .5902985      1.028401
pyperstudm~e |      .9316843      .0342739      -1.92      0.054      .8668732      1.001341
      Asian | (omitted)
      Black |      3068818      5.66e+09      0.01      0.994      0      .
      Latino |      3106495      3.98e+09      0.01      0.991      0      .
      Other |      1.39e-06      .0025095      -0.01      0.994      0      .
      AsianMale | (omitted)
      BlackMale |      1.550268      4041.894      0.00      1.000      0      .
      LatinoMale |      8.25e-07      .0010563      -0.01      0.991      0      .
      OtherMale |      1.50e+12      3.73e+15      0.01      0.991      0      .
-----

```

```

-----> region = 15

```

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note: Asian omitted because of collinearity
note: Other omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -291.40685
Iteration 1: log likelihood = -262.6446
Iteration 2: log likelihood = -260.90496
Iteration 3: log likelihood = -260.88352
Iteration 4: log likelihood = -260.88156
Iteration 5: log likelihood = -260.88116
Iteration 6: log likelihood = -260.88107
Iteration 7: log likelihood = -260.88105

```

```

Multinomial logistic regression      Number of obs   =      279
                                     LR chi2(30)        =      61.05
                                     Prob > chi2         =      0.0007
Log likelihood = -260.88105          Pseudo R2        =      0.1048

```

```

-----
      admin |      RRR      Std. Err.      z      P>|z|      [95% Conf. Interval]
-----+-----
1      |      (base outcome)
-----+-----
2      |
      tenure |      1.041088      .0337624      1.24      0.214      .9769743      1.10941
      exper |      .9384606      .0339097      -1.76      0.079      .8742977      1.007332
      male |      1.08558      .5023557      0.18      0.859      .4382942      2.688798
      age |      .9816244      .0266064      -0.68      0.494      .9308378      1.035182
pyacctrati~2 |      1.194744      .4349449      0.49      0.625      .5853233      2.438673
pyperstudlep |      .9197293      .0291263      -2.64      0.008      .8643784      .9786247
pyperstude~s |      1.010787      .0137134      0.79      0.429      .9842637      1.038026
pyperstudw~e |      1.309047      .3309518      1.07      0.287      .797545      2.148597
pyperstudb~k |      1.48456      .4107322      1.43      0.153      .8631703      2.553284
pyperstudl~o |      1.337307      .3389148      1.15      0.251      .8137861      2.197616
pyperstudm~e |      .9515027      .0347088      -1.36      0.173      .8858495      1.022022
      Asian | (omitted)
      Black |      795116.3      5.87e+08      0.02      0.985      0      .
      Latino |      3.251507      1.974903      1.94      0.052      .9887375      10.69272
      Other | (omitted)
      AsianMale | (omitted)
      BlackMale |      .0000102      .0075287      -0.02      0.988      0      .
      LatinoMale |      .8374191      .6917851      -0.21      0.830      .1658717      4.227791

```

OtherMale	(omitted)						
3							
tenure	.9785957	.0196934	-1.08	0.282	.9407486	1.017965	
exper	1.039833	.0256802	1.58	0.114	.9906998	1.091404	
male	1.061152	.3529881	0.18	0.858	.5528776	2.036697	
age	.9784215	.0216087	-0.99	0.323	.9369727	1.021704	
pyacctrati~2	1.918735	.5029963	2.49	0.013	1.147817	3.207432	
pyperstudlep	.9566258	.0244002	-1.74	0.082	.909978	1.005665	
pyperstude~s	1.023116	.011356	2.06	0.040	1.001099	1.045617	
pyperstudw~e	1.012901	.1645732	0.08	0.937	.7366586	1.392733	
pyperstudb~k	1.01493	.183706	0.08	0.935	.7118126	1.447127	
pyperstudl~o	1.010532	.1641128	0.06	0.949	.7350428	1.389271	
pyperstudm~e	.9620193	.0264018	-1.41	0.158	.9116398	1.015183	
Asian	(omitted)						
Black	1.41893	1552.627	0.00	1.000	0	.	
Latino	1.335822	.7470241	0.52	0.605	.4464121	3.997248	
Other	(omitted)						
AsianMale	(omitted)						
BlackMale	1.159918	1269.211	0.00	1.000	0	.	
LatinoMale	.7311212	.5913006	-0.39	0.699	.149821	3.567845	
OtherMale	(omitted)						

-----> region = 16

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -532.29223  
Iteration 1: log likelihood = -488.87109  
Iteration 2: log likelihood = -487.65959  
Iteration 3: log likelihood = -487.58937  
Iteration 4: log likelihood = -487.57849  
Iteration 5: log likelihood = -487.57586  
Iteration 6: log likelihood = -487.57531  
Iteration 7: log likelihood = -487.57519  
Iteration 8: log likelihood = -487.57516

Multinomial logistic regression	Number of obs	=	487
	LR chi2(34)	=	89.43
	Prob > chi2	=	0.0000
Log likelihood = -487.57516	Pseudo R2	=	0.0840

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.044911	.0204905	2.24	0.025	1.005513	1.085854
exper	.9727365	.0250452	-1.07	0.283	.9248667	1.023084
male	.7744783	.2055901	-0.96	0.336	.4603122	1.303065
age	.9670493	.0210652	-1.54	0.124	.9266313	1.00923
pyacctrati~2	1.109618	.21204	0.54	0.586	.7629829	1.613734
pyperstudlep	.997563	.0157105	-0.15	0.877	.9672414	1.028835
pyperstude~s	.981003	.0093008	-2.02	0.043	.9629421	.9994027
pyperstudw~e	.9427024	.0350728	-1.59	0.113	.8764075	1.014012
pyperstudb~k	.965647	.0398142	-0.85	0.397	.8906824	1.046921
pyperstudl~o	.9617438	.03442	-1.09	0.276	.8965935	1.031628
pyperstudm~e	1.031221	.0258157	1.23	0.219	.9818442	1.083081
Asian	(omitted)					
Black	.8027713	.7235555	-0.24	0.807	.1372098	4.696762
Latino	2.683132	1.695703	1.56	0.118	.7774887	9.259549
Other	2.208075	2.819656	0.62	0.535	.1807385	26.97595
AsianMale	(omitted)					
BlackMale	241474	1.00e+08	0.03	0.976	0	.

LatinoMale		1.303667	1.369198	0.25	0.801	.1664095	10.21305
OtherMale		1874670	2.02e+09	0.01	0.989	0	.
-----							
3							
tenure		1.024318	.0170046	1.45	0.148	.991526	1.058195
exper		1.010511	.021202	0.50	0.618	.9697986	1.052932
male		2.019125	.501469	2.83	0.005	1.240964	3.285241
age		.9998334	.0200441	-0.01	0.993	.9613094	1.039901
pyacctrati~2		1.78445	.3147703	3.28	0.001	1.262861	2.521466
pyperstudlep		1.01781	.0156664	1.15	0.251	.9875629	1.048983
pyperstude~s		1.006691	.0080679	0.83	0.405	.9910016	1.022628
pyperstudw~e		1.040047	.0471223	0.87	0.386	.9516708	1.13663
pyperstudb~k		1.021745	.0512257	0.43	0.668	.9261194	1.127244
pyperstudl~o		1.03269	.0456087	0.73	0.466	.947058	1.126064
pyperstudm~e		1.009772	.0226839	0.43	0.665	.9662768	1.055225
Asian		(omitted)					
Black		1.441472	1.417084	0.37	0.710	.2098967	9.899347
Latino		2.02103	1.447074	0.98	0.326	.4967127	8.223187
Other		2.387629	3.425524	0.61	0.544	.1434665	39.7359
AsianMale		(omitted)					
BlackMale		96441.58	4.00e+07	0.03	0.978	0	.
LatinoMale		.9328838	1.065411	-0.06	0.951	.0994725	8.748874
OtherMale		.2227468	349.432	-0.00	0.999	0	.
-----							

-----> region = 17

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -350.11603  
Iteration 1: log likelihood = -316.14231  
Iteration 2: log likelihood = -313.95523  
Iteration 3: log likelihood = -313.7195  
Iteration 4: log likelihood = -313.67608  
Iteration 5: log likelihood = -313.66615  
Iteration 6: log likelihood = -313.66396  
Iteration 7: log likelihood = -313.66341  
Iteration 8: log likelihood = -313.6633  
Iteration 9: log likelihood = -313.66328

Multinomial logistic regression	Number of obs	=	343
	LR chi2(32)	=	72.91
	Prob > chi2	=	0.0000
Log likelihood = -313.66328	Pseudo R2	=	0.1041

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.002442	.0297139	0.08	0.934	.9458635	1.062406
	exper	.9935491	.0348483	-0.18	0.854	.9275424	1.064253
	male	1.119463	.4297369	0.29	0.769	.5275362	2.375565
	age	.970201	.027634	-1.06	0.288	.9175233	1.025903
pyacctrati~2		.7768767	.2327701	-0.84	0.399	.4318298	1.397628
pyperstudlep		.9843248	.0367986	-0.42	0.673	.9147798	1.059157
pyperstude~s		.9856284	.0171386	-0.83	0.405	.9526033	1.019798
pyperstudw~e		.9460046	.096288	-0.55	0.586	.774916	1.154867
pyperstudb~k		.94904	.099311	-0.50	0.617	.7730573	1.165084
pyperstudl~o		.9441093	.0967321	-0.56	0.575	.7723415	1.154078
pyperstudm~e		1.026922	.0244948	1.11	0.265	.9800179	1.076071
	Asian	(omitted)					
	Black	5.11e-07	.0007377	-0.01	0.992	0	.
	Latino	1.406071	1.026687	0.47	0.641	.3361058	5.882185



Other		5.18e-07	.0009369	-0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		2.88e+12	4.78e+15	0.02	0.986	0	.
LatinoMale		2.708837	2.908154	0.93	0.353	.3303312	22.21345
OtherMale		(omitted)					

---

3							
tenure		1.019425	.0212491	0.92	0.356	.9786172	1.061935
exper		1.020434	.0248368	0.83	0.406	.972898	1.070293
male		2.580421	.7743794	3.16	0.002	1.433001	4.646592
age		.9918256	.0206024	-0.40	0.693	.9522566	1.033039
pyacctrati~2		1.118207	.2573192	0.49	0.627	.7122714	1.755491
pyperstudlep		1.00947	.025602	0.37	0.710	.9605177	1.060917
pyperstude~s		1.034631	.0137612	2.56	0.010	1.008008	1.061957
pyperstudw~e		1.023765	.1094518	0.22	0.826	.8302285	1.262418
pyperstudb~k		.978707	.1067458	-0.20	0.844	.7903394	1.21197
pyperstudl~o		1.017969	.1088708	0.17	0.868	.8254661	1.255364
pyperstudm~e		1.019041	.022228	0.86	0.387	.9763932	1.063552
Asian		(omitted)					
Black		1.407967	2.127784	0.23	0.821	.0728145	27.22496
Latino		.2687482	.2276945	-1.55	0.121	.0510708	1.414224
Other		1.828097	2.672927	0.41	0.680	.1040942	32.10496
AsianMale		(omitted)					
BlackMale		3947922	3.23e+09	0.02	0.985	0	.
LatinoMale		3.456792	4.067767	1.05	0.292	.3443699	34.69935
OtherMale		(omitted)					

---

-----> region = 18

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -420.3517  
Iteration 1: log likelihood = -394.36294  
Iteration 2: log likelihood = -393.8018  
Iteration 3: log likelihood = -393.76717  
Iteration 4: log likelihood = -393.76013  
Iteration 5: log likelihood = -393.75865  
Iteration 6: log likelihood = -393.75832  
Iteration 7: log likelihood = -393.75824  
Iteration 8: log likelihood = -393.75822

Multinomial logistic regression	Number of obs	=	396
	LR chi2(34)	=	53.19
	Prob > chi2	=	0.0192
Log likelihood = -393.75822	Pseudo R2	=	0.0633

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9975607	.0209492	-0.12	0.907	.9573345 1.039477
exper		.9723445	.0256824	-1.06	0.288	.9232887 1.024007
male		1.172979	.4109273	0.46	0.649	.5903241 2.330719
age		1.020861	.0225746	0.93	0.350	.977561 1.066079
pyacctrati~2		.710705	.1981095	-1.23	0.221	.4115441 1.227333
pyperstudlep		1.009648	.0186178	0.52	0.603	.9738089 1.046805
pyperstude~s		.9982567	.0122732	-0.14	0.887	.9744891 1.022604
pyperstudw~e		.9328437	.1112921	-0.58	0.560	.738341 1.178585
pyperstudb~k		.9842511	.124177	-0.13	0.900	.7686258 1.260366
pyperstudl~o		.9244977	.1082278	-0.67	0.502	.7349514 1.162929
pyperstudm~e		1.013953	.0137243	1.02	0.306	.9874077 1.041212
Asian		(omitted)				
Black		2.518994	2.460042	0.95	0.344	.3714893 17.08079

Latino		2.607161	1.034244	2.42	0.016	1.198123	5.673277
Other		.8223165	1140.578	-0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		.7418228	1.064403	-0.21	0.835	.0445611	12.34936
LatinoMale		.444497	.2682485	-1.34	0.179	.1362007	1.450636
OtherMale		4467847	7.93e+09	0.01	0.993	0	.
-----							
3							
tenure		.9878025	.0166261	-0.73	0.466	.9557476	1.020932
exper		1.070274	.0275388	2.64	0.008	1.017638	1.125634
male		1.377325	.4265977	1.03	0.301	.7505759	2.527426
age		.9759412	.0224628	-1.06	0.290	.9328932	1.020976
pyacctrati~2		1.256215	.3105056	0.92	0.356	.7738695	2.039202
pyperstudlep		.9904481	.0166876	-0.57	0.569	.9582752	1.023701
pyperstude~s		1.029733	.0118343	2.55	0.011	1.006798	1.053191
pyperstudw~e		.9700388	.1060836	-0.28	0.781	.7828914	1.201923
pyperstudb~k		.9924812	.1135702	-0.07	0.947	.7930834	1.242012
pyperstudl~o		.9584869	.1031282	-0.39	0.694	.7762495	1.183507
pyperstudm~e		.9804536	.0170567	-1.13	0.257	.9475866	1.014461
Asian		(omitted)					
Black		3.49283	3.262219	1.34	0.181	.5599839	21.78609
Latino		1.449989	.5751762	0.94	0.349	.6663667	3.155122
Other		1220986	9.90e+08	0.02	0.986	0	.
AsianMale		(omitted)					
BlackMale		.8512784	1.138502	-0.12	0.904	.0618974	11.70767
LatinoMale		.6861255	.3925542	-0.66	0.510	.2235651	2.105732
OtherMale		1.07e-06	.0020571	-0.01	0.994	0	.
-----							

-----> region = 19

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -865.06154  
Iteration 1: log likelihood = -785.71805  
Iteration 2: log likelihood = -780.67015  
Iteration 3: log likelihood = -780.38207  
Iteration 4: log likelihood = -780.31855  
Iteration 5: log likelihood = -780.30425  
Iteration 6: log likelihood = -780.30115  
Iteration 7: log likelihood = -780.30067  
Iteration 8: log likelihood = -780.30061  
Iteration 9: log likelihood = -780.3006

Multinomial logistic regression	Number of obs	=	802
	LR chi2(36)	=	169.52
	Prob > chi2	=	0.0000
Log likelihood = -780.3006	Pseudo R2	=	0.0980

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.005836	.0142411	0.41	0.681	.9783081 1.034139
exper		1.034947	.0179759	1.98	0.048	1.000308 1.070785
male		.8985258	.2806229	-0.34	0.732	.4871781 1.657194
age		.9961549	.0128086	-0.30	0.764	.9713641 1.021578
pyacctrati~2		1.003701	.1756323	0.02	0.983	.7122904 1.414333
pyperstudlep		1.001375	.0064398	0.21	0.831	.9888328 1.014077
pyperstude~s		.9958934	.0094933	-0.43	0.666	.9774597 1.014675
pyperstudw~e		.9052828	.0902239	-1.00	0.318	.7446469 1.100571
pyperstudb~k		.9496561	.0921245	-0.53	0.594	.7852226 1.148524
pyperstudl~o		.9270665	.0841549	-0.83	0.404	.775966 1.10759
pyperstudm~e		1.008755	.0137851	0.64	0.524	.9820953 1.036138
Asian		1.35675	1.963149	0.21	0.833	.0795893 23.12839

	Black		.8240017	.4094834	-0.39	0.697	.31112	2.18237
	Latino		1.218679	.3042649	0.79	0.428	.7470878	1.987957
	Other		.2548619	.294273	-1.18	0.236	.026514	2.449818
	AsianMale		(omitted)					
	BlackMale		2.411362	2.215506	0.96	0.338	.3982881	14.59915
	LatinoMale		.9119298	.3410783	-0.25	0.805	.4381235	1.898131
	OtherMale		1.93e-06	.0020819	-0.01	0.990	0	.
-----								
3	tenure		1.003074	.0154676	0.20	0.842	.9732118	1.033853
	exper		1.148241	.0255928	6.20	0.000	1.09916	1.199514
	male		.9928003	.3658432	-0.02	0.984	.4821668	2.044215
	age		.9673392	.017503	-1.84	0.066	.9336351	1.00226
	pyacctrati~2		1.26291	.2607313	1.13	0.258	.842635	1.892803
	pyperstudlep		1.032927	.0077528	4.32	0.000	1.017843	1.048234
	pyperstude~s		1.019255	.0123305	1.58	0.115	.9953717	1.043711
	pyperstudw~e		1.022677	.1206554	0.19	0.849	.8115468	1.288734
	pyperstudb~k		1.027034	.1188143	0.23	0.818	.8186764	1.288421
	pyperstudl~o		.9810235	.1062415	-0.18	0.860	.7934091	1.213002
	pyperstudm~e		.9860425	.0206786	-0.67	0.503	.946335	1.027416
	Asian		1.07e-06	.0017575	-0.01	0.993	0	.
	Black		.4290232	.3121213	-1.16	0.245	.1030904	1.785433
	Latino		.956878	.2831778	-0.15	0.882	.5357381	1.709073
	Other		2.45e-07	.0002162	-0.02	0.986	0	.
	AsianMale		(omitted)					
	BlackMale		1.131586	1.738185	0.08	0.936	.055743	22.97126
	LatinoMale		1.190703	.5333247	0.39	0.697	.4949301	2.864593
	OtherMale		1.41e+07	1.24e+10	0.02	0.985	0	.
-----								

-----> region = 20

Iteration 0: log likelihood = -2165.8402  
Iteration 1: log likelihood = -2062.2141  
Iteration 2: log likelihood = -2058.886  
Iteration 3: log likelihood = -2058.6614  
Iteration 4: log likelihood = -2058.6158  
Iteration 5: log likelihood = -2058.6063  
Iteration 6: log likelihood = -2058.6042  
Iteration 7: log likelihood = -2058.6036  
Iteration 8: log likelihood = -2058.6035  
Iteration 9: log likelihood = -2058.6035

Multinomial logistic regression	Number of obs	=	2082
	LR chi2(38)	=	214.47
	Prob > chi2	=	0.0000
Log likelihood = -2058.6035	Pseudo R2	=	0.0495

	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----							
1			(base outcome)				
-----							
2	tenure		.9993453	.0080163	-0.08	0.935	.9837566 1.015181
	exper		1.006004	.0104298	0.58	0.564	.9857679 1.026655
	male		1.29574	.198851	1.69	0.091	.9591534 1.750442
	age		.9758464	.0084743	-2.82	0.005	.9593776 .992598
	pyacctrati~2		1.036143	.1008881	0.36	0.715	.8561286 1.254008
	pyperstudlep		.9821481	.0066848	-2.65	0.008	.9691332 .9953378
	pyperstude~s		1.004109	.0052786	0.78	0.435	.9938164 1.014508
	pyperstudw~e		.9705874	.0302734	-0.96	0.338	.9130298 1.031773
	pyperstudb~k		.9658759	.0332344	-1.01	0.313	.9028856 1.033261
	pyperstudl~o		.966696	.0314576	-1.04	0.298	.9069653 1.030361
	pyperstudm~e		1.002064	.0071886	0.29	0.774	.988073 1.016253
	Asian		1.311809	1.214126	0.29	0.769	.2138226 8.047996

Black		1.425436	.4463544	1.13	0.258	.7716253	2.633232
Latino		1.495754	.2212903	2.72	0.006	1.119254	1.998903
Other		.3019241	.3247993	-1.11	0.266	.0366611	2.486511
AsianMale		1.199049	2.037461	0.11	0.915	.0428993	33.51385
BlackMale		1.752426	.8125185	1.21	0.226	.7062801	4.348128
LatinoMale		.993726	.2251445	-0.03	0.978	.637402	1.549244
OtherMale		3.736065	5.301459	0.93	0.353	.2315065	60.29282
-----							
3							
tenure		1.01364	.0082099	1.67	0.094	.997676	1.02986
exper		1.06919	.0135052	5.30	0.000	1.043046	1.09599
male		1.359523	.2317832	1.80	0.072	.9733443	1.89892
age		.9621057	.010774	-3.45	0.001	.9412191	.9834558
pyacctrati~2		1.475197	.151843	3.78	0.000	1.205689	1.804948
pyperstudlep		1.018192	.0061329	2.99	0.003	1.006243	1.030284
pyperstude~s		1.020283	.0060894	3.36	0.001	1.008418	1.032288
pyperstudw~e		1.023868	.0369197	0.65	0.513	.9540042	1.098847
pyperstudb~k		1.012878	.0399264	0.32	0.745	.9375703	1.094235
pyperstudl~o		1.003713	.0375789	0.10	0.921	.9326972	1.080136
pyperstudm~e		.9935922	.0089724	-0.71	0.477	.9761614	1.011334
Asian		2.10e-06	.0012197	-0.02	0.982	0	.
Black		1.788129	.5594195	1.86	0.063	.968498	3.301405
Latino		.9097816	.1511475	-0.57	0.569	.6569342	1.259947
Other		.4550726	.4919252	-0.73	0.466	.054695	3.786292
AsianMale		916844.1	5.33e+08	0.02	0.981	0	.
BlackMale		.8121306	.4389928	-0.38	0.700	.2815228	2.342816
LatinoMale		1.075417	.2782263	0.28	0.779	.6476775	1.785646
OtherMale		4.47e-06	.0026173	-0.02	0.983	0	.
-----							

by psychooltype:

```
. bysort psychooltype: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2008a==1, rrr iter(20)
```

```
-----> psychooltype =
no observations
```

```
-----> psychooltype = B
```

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -349.52757
Iteration 1: log likelihood = -328.41839
Iteration 2: log likelihood = -320.78775
Iteration 3: log likelihood = -319.93757
Iteration 4: log likelihood = -319.90013
Iteration 5: log likelihood = -319.8917
Iteration 6: log likelihood = -319.88971
Iteration 7: log likelihood = -319.8893
Iteration 8: log likelihood = -319.88923
Iteration 9: log likelihood = -319.88922
```

```
Multinomial logistic regression      Number of obs   =      394
                                     LR chi2(36)        =      59.28
                                     Prob > chi2         =      0.0086
Log likelihood = -319.88922          Pseudo R2        =      0.0848
```

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						

tenure		1.020861	.0319497	0.66	0.509	.9601226	1.085442
exper		1.023643	.041176	0.58	0.561	.9460391	1.107613
male		.7561549	.3451343	-0.61	0.540	.3090965	1.849812
age		.9224745	.0323744	-2.30	0.021	.8611549	.9881604
pyacctrati~2		.4443587	.1787789	-2.02	0.044	.201962	.9776824
pyperstudlep		1.006217	.0276354	0.23	0.821	.9534848	1.061866
pyperstude~s		.9810327	.0170993	-1.10	0.272	.9480848	1.015126
pyperstudw~e		.9201032	.0274781	-2.79	0.005	.8677929	.9755667
pyperstudb~k		.9432292	.0333007	-1.66	0.098	.880168	1.010809
pyperstudl~o		.951715	.029631	-1.59	0.112	.8953757	1.011599
pyperstudm~e		1.007976	.0224767	0.36	0.722	.9648716	1.053007
Asian		1.60e-07	.0004822	-0.01	0.996	0	.
Black		5.488757	5.070753	1.84	0.065	.8976289	33.56226
Latino		1.33e-06	.0019605	-0.01	0.993	0	.
Other		5.81e+09	2.85e+14	0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		.2166216	.2719515	-1.22	0.223	.0184963	2.536992
LatinoMale		1491134	2.19e+09	0.01	0.992	0	.
OtherMale		1.80e-09	.0000888	-0.00	1.000	0	.
-----							
3							
tenure		.9722903	.0167143	-1.63	0.102	.9400766	1.005608
exper		1.018767	.0192431	0.98	0.325	.9817411	1.05719
male		1.11772	.281063	0.44	0.658	.682793	1.829688
age		.9941411	.0172062	-0.34	0.734	.960983	1.028443
pyacctrati~2		.8321789	.1771406	-0.86	0.388	.5483109	1.263009
pyperstudlep		1.01309	.0177396	0.74	0.458	.9789108	1.048462
pyperstude~s		.9955444	.0090365	-0.49	0.623	.9779899	1.013414
pyperstudw~e		.9821054	.0264806	-0.67	0.503	.931552	1.035402
pyperstudb~k		.9824506	.0294278	-0.59	0.554	.9264336	1.041855
pyperstudl~o		.9808188	.0272013	-0.70	0.485	.9289283	1.035608
pyperstudm~e		1.01293	.0125154	1.04	0.298	.9886951	1.037759
Asian		5.98e-07	.0009593	-0.01	0.993	0	.
Black		1.57205	1.471522	0.48	0.629	.2510131	9.845462
Latino		1.317021	1.688203	0.21	0.830	.1067809	16.24396
Other		1.33515	111538.5	0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		.8214387	.9038331	-0.18	0.858	.0950591	7.098334
LatinoMale		2.225467	3.163274	0.56	0.574	.137254	36.08422
OtherMale		5336231	4.46e+11	0.00	1.000	0	.
-----							

-----> pyschooltype = E

Iteration 0: log likelihood = -11660.505  
Iteration 1: log likelihood = -11036.278  
Iteration 2: log likelihood = -11031.159  
Iteration 3: log likelihood = -11031.156  
Iteration 4: log likelihood = -11031.156

Multinomial logistic regression	Number of obs	=	10695
	LR chi2(38)	=	1258.70
	Prob > chi2	=	0.0000
Log likelihood = -11031.156	Pseudo R2	=	0.0540

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----							
1		(base outcome)					
-----							
2							
tenure		1.021188	.0042435	5.05	0.000	1.012904	1.029538
exper		1.001691	.0052707	0.32	0.748	.9914141	1.012075
male		1.589619	.1373548	5.36	0.000	1.341971	1.882969
age		.9764466	.003952	-5.89	0.000	.9687316	.9842231
pyacctrati~2		.9140584	.0344476	-2.38	0.017	.8489756	.9841305

pyperstudlep		1.005952	.0018105	3.30	0.001	1.002409	1.009506
pyperstude~s		.9921694	.0021336	-3.66	0.000	.9879965	.99636
pyperstudw~e		.9929213	.0051035	-1.38	0.167	.9829687	1.002975
pyperstudb~k		.9910668	.0058164	-1.53	0.126	.9797323	1.002533
pyperstudl~o		.9948203	.0054183	-0.95	0.340	.9842571	1.005497
pyperstudm~e		1.007698	.0037464	2.06	0.039	1.000382	1.015068
Asian		.3666511	.1594175	-2.31	0.021	.156371	.859706
Black		.8922106	.0840851	-1.21	0.226	.7417322	1.073217
Latino		1.09202	.0823263	1.17	0.243	.9420179	1.265907
Other		1.872788	.5048551	2.33	0.020	1.104146	3.176512
AsianMale		2.399277	1.857336	1.13	0.258	.5261938	10.93995
BlackMale		1.006195	.2069082	0.03	0.976	.6724287	1.505631
LatinoMale		.6717891	.0959852	-2.78	0.005	.5077072	.888892
OtherMale		.3442718	.2205404	-1.66	0.096	.0980894	1.208317
-----							
3							
tenure		1.024091	.003661	6.66	0.000	1.01694	1.031291
exper		1.064499	.0053196	12.51	0.000	1.054124	1.074976
male		2.512479	.1990816	11.63	0.000	2.151075	2.934602
age		.9831241	.0040769	-4.10	0.000	.9751659	.9911471
pyacctrati~2		.9748371	.0359997	-0.69	0.490	.906772	1.048011
pyperstudlep		.9972896	.0017793	-1.52	0.128	.9938083	1.000783
pyperstude~s		1.008137	.0020865	3.92	0.000	1.004056	1.012235
pyperstudw~e		.9992329	.0052984	-0.14	0.885	.988902	1.009672
pyperstudb~k		.9865258	.0059299	-2.26	0.024	.9749716	.9982169
pyperstudl~o		.9902434	.0055559	-1.75	0.081	.99794136	1.001193
pyperstudm~e		.9963159	.0037891	-0.97	0.332	.9889171	1.00377
Asian		.9269231	.3436111	-0.20	0.838	.4482317	1.916835
Black		.9080877	.0861428	-1.02	0.309	.7540173	1.09364
Latino		1.086861	.0828851	1.09	0.275	.9359666	1.262081
Other		1.221749	.3781942	0.65	0.518	.6660266	2.24116
AsianMale		1.122364	.7961288	0.16	0.871	.2794791	4.507319
BlackMale		.6283962	.1296718	-2.25	0.024	.4193581	.9416336
LatinoMale		.5455534	.0758312	-4.36	0.000	.4154525	.7163961
OtherMale		.3169513	.2222979	-1.64	0.101	.0801662	1.253124
-----							

-----> pyschooltype = M

Iteration 0: log likelihood = -6376.9921  
Iteration 1: log likelihood = -6094.9607  
Iteration 2: log likelihood = -6089.0878  
Iteration 3: log likelihood = -6089.0713  
Iteration 4: log likelihood = -6089.0713

Multinomial logistic regression	Number of obs	=	6024
	LR chi2(38)	=	575.84
	Prob > chi2	=	0.0000
Log likelihood = -6089.0713	Pseudo R2	=	0.0451

-----						
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.006782	.0049529	1.37	0.169	.9971215 1.016537
exper		1.00287	.0059716	0.48	0.630	.9912344 1.014643
male		1.625421	.1268757	6.22	0.000	1.394837 1.894123
age		.9828193	.0046422	-3.67	0.000	.9737628 .99196
pyacctrati~2		.9468979	.0529216	-0.98	0.329	.8486527 1.056517
pyperstudlep		1.004371	.003654	1.20	0.231	.9972342 1.011558
pyperstude~s		.9993062	.0027957	-0.25	0.804	.9938416 1.004801
pyperstudw~e		.9917419	.0069072	-1.19	0.234	.978296 1.005373
pyperstudb~k		.9924081	.0079041	-0.96	0.339	.9770366 1.008021
pyperstudl~o		.9905105	.00748	-1.26	0.207	.975958 1.00528

pyperstudm~e		1.004382	.004573	0.96	0.337	.9954594	1.013385
Asian		1.087993	.5012643	0.18	0.855	.4410191	2.684077
Black		1.022704	.1119215	0.21	0.837	.8252722	1.267368
Latino		2.115884	.2454629	6.46	0.000	1.68556	2.656069
Other		.7619881	.2798628	-0.74	0.459	.3709532	1.565227
AsianMale		1.537671	1.132145	0.58	0.559	.3632011	6.509982
BlackMale		1.321104	.2227547	1.65	0.099	.9493225	1.838486
LatinoMale		.7197479	.1131831	-2.09	0.037	.5288417	.9795692
OtherMale		1.861091	.9654351	1.20	0.231	.6733016	5.144293

3							
tenure		1.009807	.0051783	1.90	0.057	.999709	1.020008
exper		1.074196	.0079305	9.69	0.000	1.058764	1.089852
male		2.042543	.1781202	8.19	0.000	1.721639	2.423262
age		.9604386	.0062089	-6.24	0.000	.9483463	.9726852
pyacctrati~2		.9557217	.0623019	-0.69	0.487	.841091	1.085975
pyperstudlep		.9893798	.0048016	-2.20	0.028	.9800134	.9988356
pyperstude~s		1.012962	.0033173	3.93	0.000	1.006481	1.019484
pyperstudw~e		1.01917	.0096444	2.01	0.045	1.000441	1.038249
pyperstudb~k		1.003705	.0107608	0.34	0.730	.9828341	1.025019
pyperstudl~o		1.005091	.0101914	0.50	0.617	.9853133	1.025266
pyperstudm~e		.9936266	.0060376	-1.05	0.293	.9818634	1.005531
Asian		1.066603	.6291961	0.11	0.913	.33564	3.389473
Black		.7824088	.1160518	-1.65	0.098	.58503	1.04638
Latino		1.534357	.2263341	2.90	0.004	1.149119	2.048745
Other		.5743872	.3147192	-1.01	0.312	.1962515	1.681112
AsianMale		.8119992	.8387913	-0.20	0.840	.1072182	6.149542
BlackMale		1.44621	.3073496	1.74	0.083	.9535282	2.193458
LatinoMale		.8684611	.1646538	-0.74	0.457	.5989201	1.259308
OtherMale		1.116853	.8298719	0.15	0.882	.2603243	4.791561

-----> pyschooltype = S

Iteration 0: log likelihood = -7102.6298  
Iteration 1: log likelihood = -6811.1161  
Iteration 2: log likelihood = -6801.6771  
Iteration 3: log likelihood = -6801.6243  
Iteration 4: log likelihood = -6801.6243

Multinomial logistic regression	Number of obs	=	7357
	LR chi2(38)	=	602.01
	Prob > chi2	=	0.0000
Log likelihood = -6801.6243	Pseudo R2	=	0.0424

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.007316	.00384	1.91	0.056	.9998174 1.01487
exper		1.006235	.0049419	1.27	0.206	.9965954 1.015967
male		1.071161	.0701948	1.05	0.294	.9420507 1.217966
age		.9707023	.0041296	-6.99	0.000	.9626422 .97883
pyacctrati~2		.8450653	.0594583	-2.39	0.017	.7362075 .9700192
pyperstudlep		1.019583	.0046136	4.29	0.000	1.010581 1.028666
pyperstude~s		.9904462	.0022504	-4.22	0.000	.9860452 .9948668
pyperstudw~e		.9955815	.0059967	-0.74	0.462	.9838972 1.007405
pyperstudb~k		.9982195	.0067117	-0.27	0.791	.985151 1.011461
pyperstudl~o		.9969155	.0064087	-0.48	0.631	.9844335 1.009556
pyperstudm~e		1.003775	.003641	1.04	0.299	.9966646 1.010937
Asian		.9616895	.4106491	-0.09	0.927	.416452 2.220776
Black		1.241164	.1372412	1.95	0.051	.9993273 1.541524
Latino		1.874988	.2030676	5.80	0.000	1.516389 2.31839
Other		.707406	.2808308	-0.87	0.383	.3249025 1.540226

AsianMale		1.082084	.7898479	0.11	0.914	.2587865	4.524601
BlackMale		1.790034	.2747321	3.79	0.000	1.325012	2.418258
LatinoMale		1.205959	.1660128	1.36	0.174	.9207799	1.579463
OtherMale		1.069214	.5909035	0.12	0.904	.361948	3.158516
-----							
3							
tenure		.9964662	.0047888	-0.74	0.461	.9871244	1.005896
exper		1.064205	.0079279	8.35	0.000	1.048779	1.079857
male		1.950851	.1835135	7.10	0.000	1.622381	2.345823
age		.9585832	.0067131	-6.04	0.000	.9455156	.9718314
pyacctrati~2		.9822674	.0948554	-0.19	0.853	.8128886	1.186939
pyperstudlep		1.004047	.0071133	0.57	0.569	.990202	1.018086
pyperstude~s		1.013301	.0031404	4.26	0.000	1.007165	1.019475
pyperstudw~e		1.034128	.0113331	3.06	0.002	1.012152	1.056581
pyperstudb~k		1.021361	.0122658	1.76	0.078	.9976008	1.045686
pyperstudl~o		1.018934	.0117262	1.63	0.103	.9962085	1.042178
pyperstudm~e		.9892221	.006044	-1.77	0.076	.9774467	1.001139
Asian		.4842548	.5032342	-0.70	0.485	.06317	3.712245
Black		.8769656	.1856877	-0.62	0.535	.5790953	1.328052
Latino		1.37248	.2515146	1.73	0.084	.9583379	1.965593
Other		.6490818	.4828065	-0.58	0.561	.1510602	2.789001
AsianMale		1.321632	1.981764	0.19	0.852	.0699438	24.97308
BlackMale		1.591591	.416126	1.78	0.075	.9534161	2.656933
LatinoMale		.8997348	.1949306	-0.49	0.626	.5884355	1.37572
OtherMale		2.889944	2.413682	1.27	0.204	.5622968	14.85296

## 2008-09

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
```

```
> Male LatinoMale OtherMale if validcert2009a==1, rrr iter(20)
```

```
Iteration 0: log likelihood = -27369.807
Iteration 1: log likelihood = -26169.392
Iteration 2: log likelihood = -26143.614
Iteration 3: log likelihood = -26143.563
Iteration 4: log likelihood = -26143.563
```

```
Multinomial logistic regression      Number of obs      =      25757
                                      LR chi2(38)           =      2452.49
                                      Prob > chi2          =      0.0000
Log likelihood = -26143.563          Pseudo R2          =      0.0448
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
tenure		1.01186	.002402	4.97	0.000	1.007163 1.016579
exper		1.005672	.002945	1.93	0.053	.9999165 1.011461
male		1.139728	.0454875	3.28	0.001	1.053972 1.232461
age		.9804976	.002302	-8.39	0.000	.9759962 .9850197
pyacctrati~2		.9378764	.021675	-2.78	0.006	.8963419 .9813355
pyperstudlep		1.005848	.0011968	4.90	0.000	1.003505 1.008196
pyperstude~s		.9939578	.0012214	-4.93	0.000	.9915667 .9963546
pyperstudw~e		.9907206	.0031885	-2.90	0.004	.9844909 .9969898
pyperstudb~k		.9919398	.0036362	-2.21	0.027	.9848386 .9990923
pyperstudl~o		.9931545	.0034126	-2.00	0.046	.9864885 .9998655
pyperstudm~e		1.005197	.002108	2.47	0.013	1.001074 1.009337
Asian		.7919656	.1861775	-0.99	0.321	.4995781 1.255478
Black		1.098466	.0623944	1.65	0.098	.9827366 1.227824
Latino		1.358615	.0703716	5.92	0.000	1.227459 1.503785
Other		1.121046	.1953838	0.66	0.512	.7966544 1.577527
AsianMale		1.696182	.6778512	1.32	0.186	.775004 3.712283



BlackMale		1.79775	.1674185	6.30	0.000	1.49782	2.157739
LatinoMale		1.206363	.0926407	2.44	0.015	1.037795	1.402313
OtherMale		.9289549	.2769844	-0.25	0.805	.5178378	1.666462
-----							
3							
tenure		1.014585	.0023652	6.21	0.000	1.00996	1.019231
exper		1.059573	.0035037	17.50	0.000	1.052728	1.066463
male		1.440425	.0612336	8.58	0.000	1.325274	1.565583
age		.9785644	.0027937	-7.59	0.000	.9731042	.9840554
pyacctrati~2		1.450373	.0358803	15.03	0.000	1.381727	1.52243
pyperstudlep		1.007841	.0013024	6.04	0.000	1.005292	1.010397
pyperstude~s		1.019027	.0013715	14.00	0.000	1.016342	1.021718
pyperstudw~e		1.001998	.0037318	0.54	0.592	.9947108	1.009339
pyperstudb~k		.9874572	.0042059	-2.96	0.003	.9792481	.9957351
pyperstudl~o		.9853013	.0039136	-3.73	0.000	.9776605	.9930017
pyperstudm~e		.9958779	.0025834	-1.59	0.111	.9908274	1.000954
Asian		.7951408	.2246188	-0.81	0.417	.4570763	1.383246
Black		.8037678	.0557227	-3.15	0.002	.7016483	.9207501
Latino		1.120749	.0660843	1.93	0.053	.9984302	1.258053
Other		.863921	.1893091	-0.67	0.504	.5622793	1.327382
AsianMale		1.374759	.6701479	0.65	0.514	.5288041	3.57403
BlackMale		1.507444	.1733271	3.57	0.000	1.203287	1.888482
LatinoMale		1.029492	.0901036	0.33	0.740	.8672093	1.222144
OtherMale		1.346672	.4556914	0.88	0.379	.6937922	2.613932
-----							

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=25757)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
tenure		46.368	2	0.000
exper		341.526	2	0.000
male		73.600	2	0.000
age		99.922	2	0.000
pyacctrati~2		304.341	2	0.000
pyperstudlep		44.200	2	0.000
pyperstude~s		310.724	2	0.000
pyperstudw~e		11.315	2	0.003
pyperstudb~k		10.230	2	0.006
pyperstudl~o		14.194	2	0.001
pyperstudm~e		12.893	2	0.002
Asian		1.301	2	0.522
Black		18.023	2	0.000
Latino		34.967	2	0.000
Other		1.281	2	0.527
AsianMale		1.801	2	0.406
BlackMale		41.722	2	0.000
LatinoMale		6.477	2	0.039
OtherMale		1.111	2	0.574
-----				

\*\*\*\* Wald tests for independent variables (N=25757)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
-----				
tenure		46.318	2	0.000
exper		320.899	2	0.000
male		73.777	2	0.000
age		98.845	2	0.000
pyacctrati~2		304.078	2	0.000
pyperstudlep		43.920	2	0.000

pyperstude~s	304.743	2	0.000
pyperstudw~e	11.473	2	0.003
pyperstudb~k	10.216	2	0.006
pyperstudl~o	14.325	2	0.001
pyperstudm~e	12.838	2	0.002
Asian	1.295	2	0.523
Black	17.746	2	0.000
Latino	35.073	2	0.000
Other	1.267	2	0.531
AsianMale	1.790	2	0.409
BlackMale	41.134	2	0.000
LatinoMale	6.482	2	0.039
OtherMale	1.115	2	0.573

\*\*\*\* Hausman tests of IIA assumption (N=25757)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	chi2	df	P>chi2	evidence
2	-64.693	19	---	---
3	1.053	19	1.000	for Ho

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=25757)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-5381.727	-5373.230	16.995	20	0.653	for Ho
3	-6581.114	-6567.772	26.684	20	0.144	for Ho

\*\*\*\* Wald tests for combining alternatives (N=25757)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2-	3   1555.306	19	0.000
2-	1   406.548	19	0.000
3-	1   1554.862	19	0.000

\*\*\*\* LR tests for combining alternatives (N=25757)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2-	3   1723.906	19	0.000
2-	1   415.611	19	0.000
3-	1   1724.149	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
```

```
> AsianMale BlackMale LatinoMale OtherMale if validcert2009a==1, rrr iter(20)
```

```
-----> region = 1
```

```
note: AsianMale omitted because of collinearity
```

```
note: OtherMale omitted because of collinearity
```

```
Iteration 0: log likelihood = -1947.2799
```

```
Iteration 1: log likelihood = -1848.8205
```

```
Iteration 2: log likelihood = -1845.7607
```

```
Iteration 3: log likelihood = -1845.5012
```

```
Iteration 4: log likelihood = -1845.4601
```

```
Iteration 5: log likelihood = -1845.4504
```

```
Iteration 6: log likelihood = -1845.4482
```

```
Iteration 7: log likelihood = -1845.4477
```

```
Iteration 8: log likelihood = -1845.4476
```

```
Iteration 9: log likelihood = -1845.4476
```

```
Multinomial logistic regression
```

```
Number of obs = 1811
```

```
LR chi2(34) = 203.66
```

```
Prob > chi2 = 0.0000
```

```
Pseudo R2 = 0.0523
```

```
Log likelihood = -1845.4476
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.0171	.0089035	1.94	0.053	.9997978	1.034701
	exper	.999036	.012078	-0.08	0.936	.9756418	1.022991
	male	1.007228	.3333636	0.02	0.983	.5265076	1.926865
	age	.9685093	.0097745	-3.17	0.002	.94954	.9878577
pyacctrati~2		.7664091	.0739198	-2.76	0.006	.6343993	.9258883
pyperstudlep		1.006338	.00299	2.13	0.033	1.000495	1.012215
pyperstude~s		.9963562	.0056489	-0.64	0.520	.9853459	1.00749
pyperstudw~e		1.082802	.0853058	1.01	0.313	.9278744	1.263597
pyperstudb~k		.9861149	.1211976	-0.11	0.909	.7750172	1.254711
pyperstudl~o		1.078462	.0748301	1.09	0.276	.9413333	1.235567
pyperstudm~e		.9914071	.0074794	-1.14	0.253	.9768555	1.006175
	Asian	.474521	.4261688	-0.83	0.407	.0816185	2.758812
	Black	.6831313	.8559644	-0.30	0.761	.0586067	7.962708
	Latino	1.047838	.2295296	0.21	0.831	.682084	1.60972
	Other	1.474829	1.25925	0.46	0.649	.2766696	7.861792
	AsianMale	(omitted)					
	BlackMale	1.776886	2.708446	0.38	0.706	.0895783	35.24654
	LatinoMale	1.287768	.4548037	0.72	0.474	.6444909	2.573109
	OtherMale	(omitted)					
3							
	tenure	1.029515	.0095333	3.14	0.002	1.010999	1.048371
	exper	1.083956	.0173411	5.04	0.000	1.050496	1.118483
	male	1.077251	.3876442	0.21	0.836	.5321278	2.180812
	age	.9397499	.0134291	-4.35	0.000	.9137945	.9664426
pyacctrati~2		1.254778	.1339305	2.13	0.033	1.017917	1.546753
pyperstudlep		1.014948	.0034912	4.31	0.000	1.008129	1.021814
pyperstude~s		.9918654	.0065209	-1.24	0.214	.9791667	1.004729
pyperstudw~e		1.034149	.0877928	0.40	0.692	.8756316	1.221363
pyperstudb~k		1.054971	.1200363	0.47	0.638	.8440916	1.318535
pyperstudl~o		1.046403	.0743739	0.64	0.523	.9103306	1.202815
pyperstudm~e		.9902689	.0090452	-1.07	0.284	.9726985	1.008157
	Asian	2.71e-07	.0003173	-0.01	0.990	0	.
	Black	1.287725	1.606655	0.20	0.839	.1116363	14.85392
	Latino	.7664822	.1775541	-1.15	0.251	.486769	1.206928
	Other	.463116	.554557	-0.64	0.520	.0443018	4.841263
	AsianMale	(omitted)					
	BlackMale	6.57e-07	.0007265	-0.01	0.990	0	.

LatinoMale		1.17403	.457792	0.41	0.681	.5467243	2.5211
OtherMale		(omitted)					

-----> region = 2

```

Iteration 0: log likelihood = -767.67186
Iteration 1: log likelihood = -735.67274
Iteration 2: log likelihood = -734.61923
Iteration 3: log likelihood = -734.4441
Iteration 4: log likelihood = -734.41673
Iteration 5: log likelihood = -734.41022
Iteration 6: log likelihood = -734.4088
Iteration 7: log likelihood = -734.40847
Iteration 8: log likelihood = -734.40842

```

Multinomial logistic regression	Number of obs	=	742
	LR chi2(38)	=	66.53
	Prob > chi2	=	0.0029
Log likelihood = -734.40842	Pseudo R2	=	0.0433

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.020114	.0161935	1.25	0.210	.9888637 1.052351
exper	.9731372	.0181473	-1.46	0.144	.9382113 1.009363
male	1.07715	.3039823	0.26	0.792	.6195245 1.872809
age	.9938152	.0135642	-0.45	0.649	.9675822 1.020759
pyacctrati~2	1.049593	.1819744	0.28	0.780	.7472107 1.474345
pyperstudlep	.994572	.0215919	-0.25	0.802	.9531404 1.037805
pyperstude~s	1.007681	.0081131	0.95	0.342	.9919046 1.023709
pyperstudw~e	.9676441	.0741051	-0.43	0.668	.8327758 1.124354
pyperstudb~k	.9996383	.0852427	-0.00	0.997	.845781 1.181484
pyperstudl~o	.9638508	.0716776	-0.50	0.621	.8331237 1.115091
pyperstudm~e	.9785278	.0147647	-1.44	0.150	.9500132 1.007898
Asian	2.234306	3.25082	0.55	0.581	.1290271 38.69052
Black	1.332124	1.022193	0.37	0.709	.2960626 5.993851
Latino	1.029821	.2673302	0.11	0.910	.6191547 1.712869
Other	2.337162	2.399984	0.83	0.408	.3123252 17.48922
AsianMale	2.15e-06	.0017625	-0.02	0.987	0 .
BlackMale	.5156919	.7246013	-0.47	0.637	.032837 8.098747
LatinoMale	1.812275	.715187	1.51	0.132	.8362005 3.927697
OtherMale	4.76e-07	.0003587	-0.02	0.985	0 .
3					
tenure	1.015225	.0144042	1.06	0.287	.9873823 1.043853
exper	1.017627	.0185979	0.96	0.339	.9818205 1.054738
male	.8109506	.2446432	-0.69	0.487	.4489609 1.464806
age	.9919495	.0145192	-0.55	0.581	.9638966 1.020819
pyacctrati~2	1.398495	.2342902	2.00	0.045	1.007068 1.942061
pyperstudlep	1.016728	.0191294	0.88	0.378	.9799178 1.054921
pyperstude~s	1.018084	.0084647	2.16	0.031	1.001628 1.034811
pyperstudw~e	1.063525	.0868197	0.75	0.451	.906277 1.248058
pyperstudb~k	1.021218	.1004434	0.21	0.831	.8421656 1.23834
pyperstudl~o	1.049696	.0833643	0.61	0.541	.8983861 1.226489
pyperstudm~e	.9604522	.0164858	-2.35	0.019	.9286781 .9933136
Asian	5.10e-06	.0041552	-0.01	0.988	0 .
Black	2.13763	1.552665	1.05	0.296	.5148295 8.875682
Latino	1.158794	.2915523	0.59	0.558	.7076915 1.897441
Other	.9747074	1.232364	-0.02	0.984	.0817831 11.61675
AsianMale	745718.8	6.08e+08	0.02	0.987	0 .
BlackMale	.7039371	.9780861	-0.25	0.801	.0462201 10.72103
LatinoMale	1.941934	.7870394	1.64	0.102	.8775137 4.29749

```

OtherMale | 2.90e-06 .0020345 -0.02 0.985 0 .
-----

```

```

-----> region = 3

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -342.48577
Iteration 1: log likelihood = -298.95785
Iteration 2: log likelihood = -298.10366
Iteration 3: log likelihood = -298.09389
Iteration 4: log likelihood = -298.09149
Iteration 5: log likelihood = -298.09103
Iteration 6: log likelihood = -298.09095
Iteration 7: log likelihood = -298.09093

```

```

Multinomial logistic regression          Number of obs   =       319
                                         LR chi2(32)      =       88.79
                                         Prob > chi2      =       0.0000
Log likelihood = -298.09093              Pseudo R2       =       0.1296

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.056373	.0273657	2.12	0.034	1.004076	1.111394
	exper	.9136186	.0250713	-3.29	0.001	.8657777	.964103
	male	2.210184	.8357953	2.10	0.036	1.053273	4.63784
	age	1.012486	.0220269	0.57	0.568	.9702218	1.056592
pyacctrati~2		.9009452	.2363898	-0.40	0.691	.5387166	1.506733
pyperstudlep		.9813227	.0362097	-0.51	0.609	.9128585	1.054922
pyperstude~s		1.01673	.0203768	0.83	0.408	.9775664	1.057463
pyperstudw~e		.9878086	.0576933	-0.21	0.834	.8809637	1.107612
pyperstudb~k		.9968408	.0582058	-0.05	0.957	.8890453	1.117706
pyperstudl~o		.9974441	.0601283	-0.04	0.966	.8862906	1.122538
pyperstudm~e		1.023767	.0291251	0.83	0.409	.9682454	1.082473
	Asian	(omitted)					
	Black	1.656457	1.233266	0.68	0.498	.3849847	7.127168
	Latino	1.543736	.9832727	0.68	0.495	.4430023	5.379481
	Other	1044902	7.78e+08	0.02	0.985	0	.
	AsianMale	(omitted)					
	BlackMale	2.480506	3.593218	0.63	0.531	.1450455	42.42054
	LatinoMale	.9769341	.9079711	-0.03	0.980	.1580342	6.0392
	OtherMale	(omitted)					
3							
	tenure	1.046613	.0211223	2.26	0.024	1.006023	1.088842
	exper	.9792938	.0237046	-0.86	0.387	.9339185	1.026874
	male	2.946358	1.065997	2.99	0.003	1.449838	5.987582
	age	1.03638	.0226927	1.63	0.103	.9928444	1.081826
pyacctrati~2		1.222498	.2808583	0.87	0.382	.7792769	1.917804
pyperstudlep		1.03657	.0306082	1.22	0.224	.978282	1.098331
pyperstude~s		1.068625	.0206375	3.44	0.001	1.028932	1.109849
pyperstudw~e		1.015237	.0591387	0.26	0.795	.905699	1.138022
pyperstudb~k		.9842824	.0571876	-0.27	0.785	.8783432	1.102999
pyperstudl~o		.9636391	.0580613	-0.61	0.539	.8563035	1.084429
pyperstudm~e		1.050033	.0292723	1.75	0.080	.9941998	1.109002
	Asian	(omitted)					
	Black	.9876789	.7247926	-0.02	0.987	.2344094	4.161564
	Latino	1.616076	1.001775	0.77	0.439	.4795325	5.446349
	Other	1.462791	1649.419	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	4.528126	6.36928	1.07	0.283	.2874848	71.32175

LatinoMale		.4203664	.4275301	-0.85	0.394	.0572691	3.085569
OtherMale		(omitted)					

-----> region = 4

Iteration 0: log likelihood = -5701.9176  
Iteration 1: log likelihood = -5352.972  
Iteration 2: log likelihood = -5335.4921  
Iteration 3: log likelihood = -5335.4055  
Iteration 4: log likelihood = -5335.4055

Multinomial logistic regression	Number of obs	=	5557
	LR chi2(38)	=	733.02
	Prob > chi2	=	0.0000
Log likelihood = -5335.4055	Pseudo R2	=	0.0643

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.027789	.0053562	5.26	0.000	1.017345	1.038341
	exper	1.026159	.0068517	3.87	0.000	1.012817	1.039676
	male	1.09064	.0978187	0.97	0.333	.9148241	1.300244
	age	.968269	.0050816	-6.14	0.000	.9583604	.9782801
pyacctrati~2		.9092645	.0448393	-1.93	0.054	.8254946	1.001535
pyperstudlep		1.000171	.0027844	0.06	0.951	.9947283	1.005643
pyperstude~s		.9969186	.0033281	-0.92	0.355	.9904169	1.003463
pyperstudw~e		.9963032	.0051615	-0.71	0.475	.986238	1.006471
pyperstudb~k		.9961012	.0063446	-0.61	0.540	.9837433	1.008614
pyperstudl~o		.9972659	.0060392	-0.45	0.651	.9854992	1.009173
pyperstudm~e		1.007774	.0047727	1.64	0.102	.998463	1.017172
	Asian	.3647357	.1550847	-2.37	0.018	.1585074	.8392802
	Black	.8443748	.0830786	-1.72	0.086	.6962817	1.023966
	Latino	1.497971	.1774847	3.41	0.001	1.187545	1.889542
	Other	.9370659	.3403402	-0.18	0.858	.4598494	1.909522
	AsianMale	2.085678	1.561942	0.98	0.326	.4806085	9.051134
	BlackMale	1.924878	.3080801	4.09	0.000	1.406588	2.634144
	LatinoMale	1.044232	.2051999	0.22	0.826	.7104426	1.534846
	OtherMale	2.116392	1.386899	1.14	0.253	.5858586	7.645387
3							
	tenure	1.038575	.0060871	6.46	0.000	1.026713	1.050574
	exper	1.087138	.0095699	9.49	0.000	1.068542	1.106057
	male	1.306357	.145621	2.40	0.017	1.049972	1.625347
	age	.9619424	.0071886	-5.19	0.000	.9479556	.9761355
pyacctrati~2		1.351369	.0804147	5.06	0.000	1.202603	1.518538
pyperstudlep		1.012933	.0034531	3.77	0.000	1.006187	1.019723
pyperstude~s		1.027173	.0043777	6.29	0.000	1.018629	1.035789
pyperstudw~e		.9943032	.006225	-0.91	0.361	.9821769	1.006579
pyperstudb~k		.9790516	.0078018	-2.66	0.008	.9638792	.9944629
pyperstudl~o		.9657116	.0071441	-4.72	0.000	.9518105	.9798158
pyperstudm~e		1.000753	.0065697	0.11	0.909	.9879593	1.013713
	Asian	.4699268	.2313991	-1.53	0.125	.1790134	1.233601
	Black	.5488707	.0721934	-4.56	0.000	.4241416	.7102794
	Latino	1.181348	.1744181	1.13	0.259	.8845122	1.577801
	Other	.659584	.368799	-0.74	0.457	.2204603	1.973376
	AsianMale	3.075719	2.626026	1.32	0.188	.5770282	16.39443
	BlackMale	1.654317	.3628854	2.29	0.022	1.076222	2.542936
	LatinoMale	1.127056	.280687	0.48	0.631	.6917646	1.836253
	OtherMale	4.351827	3.785456	1.69	0.091	.791134	23.93829

-----> region = 5

Iteration 0: log likelihood = -534.18779  
 Iteration 1: log likelihood = -483.26692  
 Iteration 2: log likelihood = -481.73627  
 Iteration 3: log likelihood = -481.56596  
 Iteration 4: log likelihood = -481.53021  
 Iteration 5: log likelihood = -481.52185  
 Iteration 6: log likelihood = -481.52003  
 Iteration 7: log likelihood = -481.5198  
 Iteration 8: log likelihood = -481.51977

Multinomial logistic regression	Number of obs	=	509
	LR chi2(38)	=	105.34
	Prob > chi2	=	0.0000
Log likelihood = -481.51977	Pseudo R2	=	0.0986

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.022651	.018409	1.24	0.213	.9871987	1.059376
	exper	.9522642	.0219826	-2.12	0.034	.9101393	.9963387
	male	1.552838	.440622	1.55	0.121	.8904213	2.708052
	age	1.001627	.0185715	0.09	0.930	.9658809	1.038696
pyacctrati~2		.7368865	.136298	-1.65	0.099	.5128122	1.05887
pyperstudlep		.9836079	.0367229	-0.44	0.658	.9142026	1.058282
pyperstude~s		1.003958	.0113256	0.35	0.726	.9820036	1.026403
pyperstudw~e		1.047445	.0594695	0.82	0.414	.9371376	1.170735
pyperstudb~k		1.034407	.0611448	0.57	0.567	.9212467	1.161466
pyperstudl~o		1.068098	.0787402	0.89	0.372	.9244012	1.234132
pyperstudm~e		.9863577	.0276916	-0.49	0.625	.9335493	1.042153
	Asian	.9800195	1.79e+10	0.01	0.993	0	.
	Black	1.766912	.7347217	1.37	0.171	.782107	3.991753
	Latino	.3084792	.3505878	-1.03	0.301	.0332531	2.861671
	Other	.2038523	1.22e+09	0.02	0.981	0	.
	AsianMale	.2817973	729.3034	-0.00	1.000	0	.
	BlackMale	2.319537	1.318445	1.48	0.139	.7613266	7.066942
	LatinoMale	7.15e-06	.0057073	-0.01	0.988	0	.
	OtherMale	2.952852	5685.257	0.00	1.000	0	.
3							
	tenure	1.021947	.0163226	1.36	0.174	.9904506	1.054445
	exper	1.048766	.0269963	1.85	0.064	.9971665	1.103035
	male	2.551193	.7289061	3.28	0.001	1.457287	4.466237
	age	.9588316	.0218367	-1.85	0.065	.9169736	1.0026
pyacctrati~2		1.627983	.2842514	2.79	0.005	1.156184	2.292306
pyperstudlep		.9744193	.0371082	-0.68	0.496	.9043366	1.049933
pyperstude~s		1.014348	.0111835	1.29	0.196	.9926635	1.036505
pyperstudw~e		1.041243	.0513658	0.82	0.413	.9452819	1.146946
pyperstudb~k		1.016214	.0525753	0.31	0.756	.9182207	1.124665
pyperstudl~o		1.065336	.0679247	0.99	0.321	.9401883	1.207142
pyperstudm~e		1.010106	.0226846	0.45	0.654	.9666088	1.05556
	Asian	4.48709	13983.64	0.00	1.000	0	.
	Black	2.346049	1.029487	1.94	0.052	.9926931	5.544459
	Latino	.5233638	.6122814	-0.55	0.580	.0528429	5.183477
	Other	.1157989	6.93e+08	0.02	0.981	0	.
	AsianMale	.1996862	880.0726	-0.00	1.000	0	.
	BlackMale	1.208091	.7275019	0.31	0.754	.3711189	3.932655
	LatinoMale	1.366624	2.635316	0.16	0.871	.0312064	59.84875
	OtherMale	3.14e-07	.0009952	-0.00	0.996	0	.

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1014.5947  
 Iteration 1: log likelihood = -953.06724  
 Iteration 2: log likelihood = -951.25508  
 Iteration 3: log likelihood = -951.10848  
 Iteration 4: log likelihood = -951.08313  
 Iteration 5: log likelihood = -951.07751  
 Iteration 6: log likelihood = -951.07614  
 Iteration 7: log likelihood = -951.07585  
 Iteration 8: log likelihood = -951.07579  
 Iteration 9: log likelihood = -951.07578

Multinomial logistic regression	Number of obs	=	969
	LR chi2(36)	=	127.04
	Prob > chi2	=	0.0000
Log likelihood = -951.07578	Pseudo R2	=	0.0626

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.023832	.0132401	1.82	0.069	.9982076 1.050113
exper	1.010709	.0154867	0.70	0.487	.9808069 1.041523
male	1.304837	.237062	1.46	0.143	.913925 1.862954
age	.9844496	.0123685	-1.25	0.212	.9605039 1.008992
pyacctrati~2	1.09001	.1365695	0.69	0.492	.8526713 1.393411
pyperstudlep	1.006912	.0137393	0.50	0.614	.9803407 1.034204
pyperstude~s	.9927003	.0064673	-1.12	0.261	.9801052 1.005457
pyperstudw~e	1.032623	.0300284	1.10	0.270	.975414 1.093187
pyperstudb~k	1.028134	.0321671	0.89	0.375	.9669817 1.093153
pyperstudl~o	1.043727	.0310338	1.44	0.150	.9846401 1.106359
pyperstudm~e	.9882669	.0117061	-1.00	0.319	.9655876 1.011479
Asian	1.428387	2.047772	0.25	0.804	.0860086 23.72193
Black	1.933973	.7344331	1.74	0.082	.9187632 4.070963
Latino	.3189079	.182465	-2.00	0.046	.1039071 .9787802
Other	.6686175	.7912469	-0.34	0.734	.0657445 6.799796
AsianMale	(omitted)				
BlackMale	2.593153	1.87774	1.32	0.188	.627279 10.72002
LatinoMale	3.496941	2.688931	1.63	0.104	.7747596 15.78373
OtherMale	9285557	2.00e+10	0.01	0.994	0 .
3					
tenure	1.016867	.0132052	1.29	0.198	.9913116 1.04308
exper	1.059413	.018166	3.37	0.001	1.0244 1.095622
male	2.210281	.4352782	4.03	0.000	1.50251 3.251453
age	.9867006	.0146214	-0.90	0.366	.9584553 1.015778
pyacctrati~2	1.693223	.238385	3.74	0.000	1.284918 2.231274
pyperstudlep	1.013024	.0151089	0.87	0.386	.9838395 1.043074
pyperstude~s	1.019065	.0070347	2.74	0.006	1.00537 1.032946
pyperstudw~e	1.045607	.0332283	1.40	0.161	.982468 1.112805
pyperstudb~k	1.045198	.0350517	1.32	0.187	.9787066 1.116206
pyperstudl~o	1.036618	.0335523	1.11	0.267	.9728996 1.104511
pyperstudm~e	.9872181	.0111686	-1.14	0.255	.9655689 1.009353
Asian	1.70e-06	.0019521	-0.01	0.991	0 .
Black	.8891915	.4429782	-0.24	0.814	.3349214 2.360737
Latino	.3353441	.2207649	-1.66	0.097	.0922837 1.218587
Other	7.50e-07	.0006589	-0.02	0.987	0 .
AsianMale	(omitted)				
BlackMale	2.615654	2.222389	1.13	0.258	.4947179 13.82939
LatinoMale	1.452824	1.383177	0.39	0.695	.2248108 9.388775
OtherMale	382410.6	1.49e+09	0.00	0.997	0 .



-----> region = 7

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -1143.0591  
 Iteration 1: log likelihood = -1072.2633  
 Iteration 2: log likelihood = -1070.877  
 Iteration 3: log likelihood = -1070.5385  
 Iteration 4: log likelihood = -1070.5021  
 Iteration 5: log likelihood = -1070.4996  
 Iteration 6: log likelihood = -1070.4991  
 Iteration 7: log likelihood = -1070.499  
 Iteration 8: log likelihood = -1070.499

Multinomial logistic regression	Number of obs	=	1100
	LR chi2(34)	=	145.12
	Prob > chi2	=	0.0000
Log likelihood = -1070.499	Pseudo R2	=	0.0635

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9958404	.0119128	-0.35	0.728	.9727632	1.019465
	exper	.9934083	.014257	-0.46	0.645	.9658545	1.021748
	male	1.991685	.370776	3.70	0.000	1.382799	2.868679
	age	.9810959	.0118976	-1.57	0.116	.9580519	1.004694
pyacctrati~2		.9916354	.1518821	-0.05	0.956	.7344787	1.338828
pyperstudlep		1.008026	.0132233	0.61	0.542	.9824394	1.034279
pyperstude~s		1.001241	.0087281	0.14	0.887	.9842792	1.018495
pyperstudw~e		.9003381	.0614581	-1.54	0.124	.7875926	1.029223
pyperstudb~k		.9070562	.0634707	-1.39	0.163	.7908095	1.040391
pyperstudl~o		.904634	.0639569	-1.42	0.156	.7875781	1.039088
pyperstudm~e		.9929838	.0161719	-0.43	0.666	.9617881	1.025191
	Asian	(omitted)					
	Black	1.910513	.6091584	2.03	0.042	1.0227	3.569042
	Latino	1.107543	.7971661	0.14	0.887	.2702086	4.539647
	Other	1943743	9.75e+08	0.03	0.977	0	.
	AsianMale	(omitted)					
	BlackMale	1.148244	.5806784	0.27	0.785	.4261605	3.093822
	LatinoMale	3.859596	4.211253	1.24	0.216	.4547632	32.75656
	OtherMale	8.53e-07	.0004277	-0.03	0.978	0	.
3							
	tenure	.9830519	.0100774	-1.67	0.095	.9634977	1.003003
	exper	1.030847	.0138487	2.26	0.024	1.004059	1.058351
	male	2.262137	.3934483	4.69	0.000	1.608687	3.181021
	age	.9882719	.0118834	-0.98	0.327	.9652533	1.01184
pyacctrati~2		1.593294	.2054799	3.61	0.000	1.237429	2.0515
pyperstudlep		1.023743	.0132044	1.82	0.069	.9981871	1.049953
pyperstude~s		1.032752	.0079956	4.16	0.000	1.017199	1.048542
pyperstudw~e		.9360063	.0708174	-0.87	0.382	.8070076	1.085625
pyperstudb~k		.9213491	.0711449	-1.06	0.289	.7919469	1.071895
pyperstudl~o		.9101148	.070786	-1.21	0.226	.781434	1.059986
pyperstudm~e		.9789176	.0144336	-1.45	0.148	.9510331	1.00762
	Asian	(omitted)					
	Black	1.915368	.6226991	2.00	0.046	1.012793	3.622293
	Latino	.6542969	.5451702	-0.51	0.611	.1278029	3.349724
	Other	361775.3	1.81e+08	0.03	0.980	0	.
	AsianMale	(omitted)					
	BlackMale	1.282063	.638944	0.50	0.618	.482719	3.405055
	LatinoMale	5.240263	6.338945	1.37	0.171	.4894264	56.10724
	OtherMale	8.63e-12	6.63e-09	-0.03	0.974	0	.

-----> region = 8

note: AsianMale omitted because of collinearity  
 note: LatinoMale omitted because of collinearity  
 Iteration 0: log likelihood = -404.57753  
 Iteration 1: log likelihood = -362.63957  
 Iteration 2: log likelihood = -361.17093  
 Iteration 3: log likelihood = -361.00247  
 Iteration 4: log likelihood = -360.96854  
 Iteration 5: log likelihood = -360.96135  
 Iteration 6: log likelihood = -360.95975  
 Iteration 7: log likelihood = -360.95936  
 Iteration 8: log likelihood = -360.95928  
 Iteration 9: log likelihood = -360.95926

Multinomial logistic regression	Number of obs	=	386
	LR chi2(34)	=	87.24
	Prob > chi2	=	0.0000
Log likelihood = -360.95926	Pseudo R2	=	0.1078

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.039833	.0217335	1.87	0.062	.9980964	1.083314
	exper	.997068	.0282097	-0.10	0.917	.943283	1.05392
	male	3.096367	.9659679	3.62	0.000	1.679982	5.7069
	age	.9723437	.0237938	-1.15	0.252	.9268094	1.020115
pyacctrati~2		1.044688	.2490599	0.18	0.855	.6547191	1.666933
pyperstudlep		1.00339	.0286508	0.12	0.906	.9487783	1.061146
pyperstude~s		.9699712	.0136469	-2.17	0.030	.9435891	.9970909
pyperstudw~e		1.046827	.149223	0.32	0.748	.7916591	1.384242
pyperstudb~k		1.084034	.1555025	0.56	0.574	.8183517	1.435973
pyperstudl~o		1.077271	.1557896	0.51	0.607	.8113873	1.430281
pyperstudm~e		1.059325	.0442541	1.38	0.168	.9760445	1.149711
	Asian	.8481954	3121.093	-0.00	1.000	0	.
	Black	2.303838	1.235123	1.56	0.120	.8055814	6.588618
	Latino	4722674	2.89e+09	0.03	0.980	0	.
	Other	9.528798	12.35322	1.74	0.082	.7508143	120.9327
	AsianMale	(omitted)					
	BlackMale	1.02366	1.339432	0.02	0.986	.0787735	13.30242
	LatinoMale	(omitted)					
	OtherMale	2.54e-08	.0000306	-0.01	0.988	0	.
3							
	tenure	1.015824	.0186241	0.86	0.392	.9799695	1.05299
	exper	1.004501	.0256007	0.18	0.860	.9555575	1.055952
	male	2.039074	.5786144	2.51	0.012	1.169212	3.55609
	age	1.010116	.0229351	0.44	0.658	.9661497	1.056083
pyacctrati~2		1.505513	.3049403	2.02	0.043	1.012219	2.239209
pyperstudlep		1.015945	.0276516	0.58	0.561	.9631694	1.071613
pyperstude~s		1.03881	.0136359	2.90	0.004	1.012425	1.065882
pyperstudw~e		.9926287	.1214637	-0.06	0.952	.7809603	1.261667
pyperstudb~k		.9741496	.1204894	-0.21	0.832	.7644399	1.241389
pyperstudl~o		.9667018	.1201141	-0.27	0.785	.7577562	1.233263
pyperstudm~e		.9906346	.0347773	-0.27	0.789	.9247645	1.061197
	Asian	2383120	5.02e+09	0.01	0.994	0	.
	Black	1.259343	.7422737	0.39	0.696	.3966747	3.9981
	Latino	1381133	8.44e+08	0.02	0.982	0	.
	Other	3.321926	4.343739	0.92	0.359	.2560717	43.09414
	AsianMale	(omitted)					
	BlackMale	2.595617	3.608898	0.69	0.493	.1701162	39.60369
	LatinoMale	(omitted)					
	OtherMale	2.20e-07	.0002551	-0.01	0.989	0	.

-----> region = 9

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -226.62976  
 Iteration 1: log likelihood = -196.83376  
 Iteration 2: log likelihood = -195.37379  
 Iteration 3: log likelihood = -194.85199  
 Iteration 4: log likelihood = -194.7893  
 Iteration 5: log likelihood = -194.77363  
 Iteration 6: log likelihood = -194.77013  
 Iteration 7: log likelihood = -194.76932  
 Iteration 8: log likelihood = -194.76915  
 Iteration 9: log likelihood = -194.76912  
 Iteration 10: log likelihood = -194.76912

Multinomial logistic regression	Number of obs	=	211
	LR chi2(36)	=	63.72
	Prob > chi2	=	0.0030
Log likelihood = -194.76912	Pseudo R2	=	0.1406

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9947532	.0325906	-0.16	0.872	.9328845	1.060725
	exper	.9858982	.0413161	-0.34	0.735	.9081565	1.070295
	male	.6033021	.2718243	-1.12	0.262	.2494691	1.458992
	age	.9783468	.0331489	-0.65	0.518	.9154865	1.045523
pyacctrati~2		2.130803	.8288964	1.94	0.052	.9940779	4.567371
pyperstudlep		1.058388	.0698914	0.86	0.390	.9298975	1.204632
pyperstude~s		.9944895	.0193975	-0.28	0.777	.9571886	1.033244
pyperstudw~e		.7045559	.0954516	-2.58	0.010	.5402523	.918828
pyperstudb~k		.6973079	.1009624	-2.49	0.013	.5250252	.9261239
pyperstudl~o		.7200391	.0979458	-2.41	0.016	.5515291	.9400342
pyperstudm~e		.9832911	.0430404	-0.38	0.700	.9024508	1.071373
	Asian	5.07e+08	4.95e+12	0.00	0.998	0	.
	Black	4.241806	5.933359	1.03	0.302	.2734676	65.79541
	Latino	3.12e-08	.0002018	-0.00	0.998	0	.
	Other	1.361508	2.028201	0.21	0.836	.073453	25.23663
	AsianMale	(omitted)					
	BlackMale	1.963406	3.713499	0.36	0.721	.0482042	79.97146
	LatinoMale	2.36e+07	2.01e+11	0.00	0.998	0	.
	OtherMale	2.28e+07	6.43e+10	0.01	0.995	0	.
3							
	tenure	1.00647	.0242396	0.27	0.789	.9600649	1.055118
	exper	1.034267	.0341759	1.02	0.308	.9694063	1.103467
	male	2.118329	.7891906	2.01	0.044	1.020646	4.396544
	age	.9814333	.0298649	-0.62	0.538	.9246105	1.041748
pyacctrati~2		2.547139	.8600256	2.77	0.006	1.314165	4.936914
pyperstudlep		1.020488	.0606557	0.34	0.733	.9082681	1.146572
pyperstude~s		1.00153	.0158554	0.10	0.923	.970931	1.033093
pyperstudw~e		1.064385	.1184489	0.56	0.575	.8558024	1.323804
pyperstudb~k		1.049597	.1232695	0.41	0.680	.8337843	1.32127
pyperstudl~o		1.072702	.120686	0.62	0.533	.8604255	1.337349
pyperstudm~e		1.010784	.0353851	0.31	0.759	.9437567	1.082573
	Asian	.9245232	12844.19	-0.00	1.000	0	.
	Black	2.94845	4.49477	0.71	0.478	.1485861	58.50721
	Latino	9.40e-08	.0004761	-0.00	0.997	0	.
	Other	2.94e-07	.0008248	-0.01	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	3.18e-08	.0001064	-0.01	0.996	0	.

LatinoMale		1.65e+14	1.03e+18	0.01	0.996	0	.
OtherMale		3.68e+13	1.46e+17	0.01	0.994	0	.

-----> region = 10

Iteration 0: log likelihood = -4184.635  
Iteration 1: log likelihood = -3923.0097  
Iteration 2: log likelihood = -3915.8349  
Iteration 3: log likelihood = -3915.8016  
Iteration 4: log likelihood = -3915.8016

Multinomial logistic regression	Number of obs	=	3949
	LR chi2(38)	=	537.67
	Prob > chi2	=	0.0000
Log likelihood = -3915.8016	Pseudo R2	=	0.0642

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.012255	.0064434	1.91	0.056	.9997043	1.024963
	exper	1.035686	.0079586	4.56	0.000	1.020204	1.051402
	male	1.263624	.1298366	2.28	0.023	1.033136	1.545533
	age	.9613458	.0059647	-6.35	0.000	.9497261	.9731077
pyacctrati~2		.9350626	.0531131	-1.18	0.237	.8365484	1.045178
pyperstudlep		.9976922	.0040071	-0.58	0.565	.9898692	1.005577
pyperstude~s		.9895051	.0036329	-2.87	0.004	.9824102	.9966512
pyperstudw~e		.9894652	.0069715	-1.50	0.133	.9758952	1.003224
pyperstudb~k		.9890068	.0078492	-1.39	0.164	.9737416	1.004511
pyperstudl~o		.9967232	.0077889	-0.42	0.674	.9815735	1.012107
pyperstudm~e		1.009895	.0061043	1.63	0.103	.9980016	1.02193
	Asian	1.550807	.8231954	0.83	0.408	.5479298	4.389251
	Black	1.274439	.1556836	1.99	0.047	1.003084	1.6192
	Latino	1.835121	.2972105	3.75	0.000	1.336001	2.520709
	Other	1.102022	.4509252	0.24	0.812	.4941919	2.45745
	AsianMale	.8650838	.6886984	-0.18	0.856	.1817237	4.118176
	BlackMale	1.552638	.3063964	2.23	0.026	1.054617	2.285839
	LatinoMale	1.081433	.2824225	0.30	0.764	.6481919	1.804247
	OtherMale	1.339512	.9149624	0.43	0.669	.3511785	5.109347
3							
	tenure	1.023125	.0065382	3.58	0.000	1.01039	1.03602
	exper	1.078667	.0094964	8.60	0.000	1.060214	1.097441
	male	1.386558	.1580676	2.87	0.004	1.108922	1.733704
	age	.9652008	.007297	-4.68	0.000	.9510043	.9796092
pyacctrati~2		1.516129	.0957896	6.59	0.000	1.339543	1.715992
pyperstudlep		1.017124	.0047959	3.60	0.000	1.007767	1.026567
pyperstude~s		1.016082	.00406	3.99	0.000	1.008155	1.02407
pyperstudw~e		1.002659	.0076969	0.35	0.729	.9876862	1.017859
pyperstudb~k		.987997	.0088038	-1.36	0.175	.9708917	1.005404
pyperstudl~o		.9754553	.0082885	-2.92	0.003	.9593447	.9918365
pyperstudm~e		1.004829	.0077392	0.63	0.532	.9897744	1.020113
	Asian	1.722372	1.10515	0.85	0.397	.4897322	6.057527
	Black	.8115995	.1234802	-1.37	0.170	.6023326	1.093572
	Latino	1.730119	.3149392	3.01	0.003	1.210958	2.471855
	Other	.6863677	.3705239	-0.70	0.486	.2382595	1.977258
	AsianMale	.2647262	.3375589	-1.04	0.297	.0217475	3.222442
	BlackMale	1.657947	.4091204	2.05	0.040	1.022176	2.689156
	LatinoMale	1.186309	.3504529	0.58	0.563	.6648756	2.116679
	OtherMale	1.687967	1.385778	0.64	0.524	.3377143	8.436817

-----> region = 11

Iteration 0: log likelihood = -3027.7975  
 Iteration 1: log likelihood = -2850.9061  
 Iteration 2: log likelihood = -2845.95  
 Iteration 3: log likelihood = -2845.931  
 Iteration 4: log likelihood = -2845.931

Multinomial logistic regression	Number of obs	=	2829
	LR chi2(38)	=	363.73
	Prob > chi2	=	0.0000
Log likelihood = -2845.931	Pseudo R2	=	0.0601

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9987194	.0075627	-0.17	0.866	.9840063	1.013652
	exper	1.020041	.0091073	2.22	0.026	1.002347	1.038048
	male	1.132402	.1210774	1.16	0.245	.9183106	1.396407
	age	.974202	.0069598	-3.66	0.000	.9606561	.9879389
pyacctrati~2		1.063492	.0758538	0.86	0.388	.924745	1.223056
pyperstudlep		1.022035	.0055027	4.05	0.000	1.011307	1.032877
pyperstude~s		.9973109	.0041155	-0.65	0.514	.9892771	1.00541
pyperstudw~e		1.000438	.0101663	0.04	0.966	.9807096	1.020563
pyperstudb~k		.9959005	.0116243	-0.35	0.725	.9733759	1.018946
pyperstudl~o		.990236	.0109037	-0.89	0.373	.9690941	1.011839
pyperstudm~e		1.005868	.0065116	0.90	0.366	.9931857	1.018711
	Asian	1.973609	1.352026	0.99	0.321	.5153989	7.557514
	Black	1.505714	.2608832	2.36	0.018	1.072164	2.114576
	Latino	1.392148	.306917	1.50	0.133	.9037055	2.144587
	Other	.6820335	.2853861	-0.91	0.360	.3003544	1.548736
	AsianMale	.6003962	.6432699	-0.48	0.634	.073528	4.90256
	BlackMale	2.227425	.666144	2.68	0.007	1.239478	4.002832
	LatinoMale	1.304484	.4489042	0.77	0.440	.6645378	2.560693
	OtherMale	1.039268	.7778293	0.05	0.959	.2396927	4.506096
3							
	tenure	1.00416	.0075767	0.55	0.582	.9894189	1.01912
	exper	1.085962	.0112705	7.95	0.000	1.064095	1.108278
	male	1.533822	.1847639	3.55	0.000	1.211265	1.942274
	age	.9754714	.0086819	-2.79	0.005	.9586028	.9926368
pyacctrati~2		1.678979	.1359916	6.40	0.000	1.432521	1.96784
pyperstudlep		1.022252	.0064188	3.51	0.000	1.009749	1.034911
pyperstude~s		1.027055	.004926	5.57	0.000	1.017446	1.036756
pyperstudw~e		1.013658	.012241	1.12	0.261	.989948	1.037936
pyperstudb~k		1.003066	.0139814	0.22	0.826	.9760341	1.030847
pyperstudl~o		.9867969	.0129293	-1.01	0.310	.9617786	1.012466
pyperstudm~e		.9797906	.0093582	-2.14	0.033	.9616194	.9983052
	Asian	.9115647	.8166719	-0.10	0.918	.157469	5.276913
	Black	1.150253	.2570651	0.63	0.531	.74227	1.78248
	Latino	1.231811	.3232171	0.79	0.427	.7365387	2.060119
	Other	.6749972	.3570278	-0.74	0.457	.2393723	1.903399
	AsianMale	1.394074	1.82829	0.25	0.800	.1066493	18.22273
	BlackMale	1.709499	.6612658	1.39	0.166	.8009554	3.648626
	LatinoMale	1.412332	.5691899	0.86	0.392	.6410475	3.111595
	OtherMale	1.679585	1.374935	0.63	0.526	.3375943	8.356205

-----> region = 12

note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -975.46935  
 Iteration 1: log likelihood = -906.47881

```

Iteration 2: log likelihood = -904.9292
Iteration 3: log likelihood = -904.76597
Iteration 4: log likelihood = -904.73629
Iteration 5: log likelihood = -904.72878
Iteration 6: log likelihood = -904.72724
Iteration 7: log likelihood = -904.7269
Iteration 8: log likelihood = -904.72683
Iteration 9: log likelihood = -904.72681

```

```

Multinomial logistic regression
Number of obs   =      906
LR chi2(36)     =     141.49
Prob > chi2     =     0.0000
Pseudo R2      =     0.0725

Log likelihood = -904.72681

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.004289	.0122452	0.35	0.726	.9805736	1.028579
	exper	1.013494	.0159093	0.85	0.393	.9827871	1.04516
	male	.8867686	.1690762	-0.63	0.529	.6102608	1.28856
	age	.9781301	.0124646	-1.74	0.083	.9540024	1.002868
pyacctrati~2		1.15102	.1647504	0.98	0.326	.8694541	1.523768
pyperstudlep		1.006074	.0160903	0.38	0.705	.9750271	1.03811
pyperstude~s		1.001232	.0077202	0.16	0.873	.9862149	1.016479
pyperstudw~e		.9358343	.0526352	-1.18	0.238	.8381541	1.044898
pyperstudb~k		.9548166	.0564395	-0.78	0.434	.8503646	1.072099
pyperstudl~o		.9355405	.0527051	-1.18	0.237	.8377391	1.04476
pyperstudm~e		1.013806	.0130255	1.07	0.286	.9885953	1.03966
	Asian	6513665	8.87e+09	0.01	0.991	0	.
	Black	.6771458	.2060185	-1.28	0.200	.3729993	1.229296
	Latino	.3980571	.1895353	-1.93	0.053	.156547	1.012152
	Other	1.238134	2375.43	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	2.855062	1.491512	2.01	0.045	1.025509	7.948616
	LatinoMale	4.333226	3.107239	2.04	0.041	1.062766	17.6679
	OtherMale	2.82e-06	.0059637	-0.01	0.995	0	.
3							
	tenure	.9971333	.0115535	-0.25	0.804	.9747441	1.020037
	exper	1.057252	.017596	3.35	0.001	1.023321	1.092308
	male	.9827416	.1906461	-0.09	0.928	.6719103	1.437366
	age	.970968	.0138659	-2.06	0.039	.9441682	.9985286
pyacctrati~2		1.508591	.2132653	2.91	0.004	1.143509	1.990231
pyperstudlep		1.05382	.0173972	3.18	0.001	1.020268	1.088476
pyperstude~s		1.01649	.008579	1.94	0.053	.9998133	1.033444
pyperstudw~e		1.055376	.0685103	0.83	0.406	.9292897	1.19857
pyperstudb~k		1.059003	.071994	0.84	0.399	.926894	1.209941
pyperstudl~o		1.018294	.0668834	0.28	0.783	.8952921	1.158195
pyperstudm~e		.9824544	.0155518	-1.12	0.263	.9524414	1.013413
	Asian	1.521186	3333.954	0.00	1.000	0	.
	Black	.5792912	.2202819	-1.44	0.151	.2749275	1.220607
	Latino	.2639256	.1729681	-2.03	0.042	.073052	.9535223
	Other	1.15e+07	1.43e+10	0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	4.623413	2.695148	2.63	0.009	1.474917	14.49298
	LatinoMale	7.109468	6.264922	2.23	0.026	1.264012	39.98738
	OtherMale	4.08e-07	.0005061	-0.01	0.991	0	.

```

-----> region = 13

```

```

Iteration 0: log likelihood = -1751.1751
Iteration 1: log likelihood = -1657.5045

```

```

Iteration 2: log likelihood = -1655.853
Iteration 3: log likelihood = -1655.7971
Iteration 4: log likelihood = -1655.7883
Iteration 5: log likelihood = -1655.7862
Iteration 6: log likelihood = -1655.7858
Iteration 7: log likelihood = -1655.7857
Iteration 8: log likelihood = -1655.7856

```

Multinomial logistic regression

```

Number of obs   =      1609
LR chi2(38)     =      190.78
Prob > chi2     =      0.0000
Pseudo R2      =      0.0545

```

Log likelihood = -1655.7856

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1							
	tenure	1.017763	.0105396	1.70	0.089	.9973139	1.038631
	exper	1.023271	.0106049	2.22	0.026	1.002695	1.044269
	male	1.027443	.1547278	0.18	0.857	.7648409	1.380208
	age	1.007706	.0083583	0.93	0.355	.9914562	1.024222
	pyacctrati~2	.942233	.0888541	-0.63	0.528	.7832288	1.133517
	pyperstudlep	1.0115	.0065983	1.75	0.080	.9986499	1.024516
	pyperstude~s	.9998981	.0066682	-0.02	0.988	.9869138	1.013053
	pyperstudw~e	1.040072	.0164484	2.48	0.013	1.008328	1.072815
	pyperstudb~k	1.04594	.0203995	2.30	0.021	1.006712	1.086696
	pyperstudl~o	1.031567	.0174496	1.84	0.066	.9979275	1.066341
	pyperstudm~e	.9777195	.0105468	-2.09	0.037	.9572651	.9986109
	Asian	.4673835	.5818247	-0.61	0.541	.0407428	5.361618
	Black	.7627893	.1898177	-1.09	0.277	.4683665	1.242291
	Latino	.7922674	.1687733	-1.09	0.274	.521848	1.202817
	Other	1.030446	.8563657	0.04	0.971	.2021264	5.253242
	AsianMale	1136198	7.75e+08	0.02	0.984	0	.
	BlackMale	.6039362	.2903943	-1.05	0.294	.2353437	1.549814
	LatinoMale	.4730928	.1728718	-2.05	0.041	.2311589	.968238
	OtherMale	1.73886	1.920449	0.50	0.616	.1996054	15.14806
2		(base outcome)					
3							
	tenure	1.02359	.0109911	2.17	0.030	1.002273	1.04536
	exper	1.054818	.0122448	4.60	0.000	1.031089	1.079092
	male	.9416692	.1632279	-0.35	0.729	.6704279	1.322649
	age	1.003477	.0096606	0.36	0.718	.98472	1.022591
	pyacctrati~2	1.478722	.1514186	3.82	0.000	1.209832	1.807374
	pyperstudlep	1.003751	.0070368	0.53	0.593	.9900537	1.017638
	pyperstude~s	1.019554	.0073934	2.67	0.008	1.005166	1.034149
	pyperstudw~e	1.013957	.016664	0.84	0.399	.9818167	1.04715
	pyperstudb~k	1.006955	.0207744	0.34	0.737	.9670502	1.048506
	pyperstudl~o	.9974302	.0175653	-0.15	0.884	.9635901	1.032459
	pyperstudm~e	.9846445	.0106467	-1.43	0.152	.9639969	1.005734
	Asian	2.864083	2.569887	1.17	0.241	.4934221	16.62465
	Black	.7099927	.207397	-1.17	0.241	.4005052	1.258634
	Latino	.8728402	.2047723	-0.58	0.562	.5511108	1.38239
	Other	2.050053	1.598997	0.92	0.357	.4444738	9.455486
	AsianMale	536183.9	3.66e+08	0.02	0.985	0	.
	BlackMale	.4957795	.3229864	-1.08	0.281	.1382785	1.777552
	LatinoMale	1.358334	.4854862	0.86	0.392	.6741807	2.736762
	OtherMale	.567072	.6915922	-0.47	0.642	.0519429	6.190852

-----> region = 14

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: BlackMale omitted because of collinearity

```

```

Iteration 0: log likelihood = -233.85337
Iteration 1: log likelihood = -200.47233
Iteration 2: log likelihood = -197.89815
Iteration 3: log likelihood = -197.74345
Iteration 4: log likelihood = -197.70692
Iteration 5: log likelihood = -197.69964
Iteration 6: log likelihood = -197.69809
Iteration 7: log likelihood = -197.69773
Iteration 8: log likelihood = -197.69765
Iteration 9: log likelihood = -197.69763

```

```

Multinomial logistic regression      Number of obs   =      244
                                     LR chi2(32)        =      72.31
                                     Prob > chi2         =      0.0001
Log likelihood = -197.69763          Pseudo R2        =      0.1546

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.019335	.0330014	0.59	0.554	.9566629	1.086113
	exper	.9637525	.0397176	-0.90	0.370	.8889684	1.044828
	male	1.247065	.6134862	0.45	0.654	.4754949	3.270639
	age	1.053991	.0373629	1.48	0.138	.9832472	1.129825
pyacctrati~2		.7116031	.2688178	-0.90	0.368	.3393789	1.492076
pyperstudlep		.8939371	.1207452	-0.83	0.406	.6860157	1.164876
pyperstude~s		.9976973	.0225076	-0.10	0.919	.9545443	1.042801
pyperstudw~e		.746876	.1866568	-1.17	0.243	.4576348	1.218928
pyperstudb~k		.7283939	.2039845	-1.13	0.258	.4207169	1.26108
pyperstudl~o		.7614959	.1893489	-1.10	0.273	.4677492	1.239716
pyperstudm~e		.9833166	.0638095	-0.26	0.795	.8658787	1.116682
	Asian	(omitted)					
	Black	1.457888	9099.936	0.00	1.000	0	.
	Latino	2.306985	8637.773	0.00	1.000	0	.
	Other	5.94e-07	.0028378	-0.00	0.998	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	2.736539	10246.1	0.00	1.000	0	.
	OtherMale	2390749	1.55e+10	0.00	0.998	0	.
3							
	tenure	.9987342	.0224264	-0.06	0.955	.9557325	1.043671
	exper	1.051824	.0340644	1.56	0.119	.9871339	1.120754
	male	2.623662	.9376259	2.70	0.007	1.302301	5.285722
	age	1.001251	.0276006	0.05	0.964	.9485902	1.056835
pyacctrati~2		1.93496	.440126	2.90	0.004	1.238961	3.021943
pyperstudlep		1.04848	.0634959	0.78	0.434	.9311322	1.180617
pyperstude~s		1.060976	.0162295	3.87	0.000	1.029639	1.093267
pyperstudw~e		.9662443	.1796784	-0.18	0.853	.6711218	1.391145
pyperstudb~k		.9116257	.1897653	-0.44	0.657	.6062186	1.370894
pyperstudl~o		.9486331	.1758485	-0.28	0.776	.6596456	1.364225
pyperstudm~e		.976587	.0250707	-0.92	0.356	.928665	1.026982
	Asian	(omitted)					
	Black	8878866	2.37e+10	0.01	0.995	0	.
	Latino	9864423	1.73e+10	0.01	0.993	0	.
	Other	1.77e-07	.0004478	-0.01	0.995	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	3.10e-07	.0005445	-0.01	0.993	0	.
	OtherMale	9.20e+13	2.89e+17	0.01	0.992	0	.

```

-----> region = 15

```



note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -288.45687  
Iteration 1: log likelihood = -260.90471  
Iteration 2: log likelihood = -259.21963  
Iteration 3: log likelihood = -258.94688  
Iteration 4: log likelihood = -258.93387  
Iteration 5: log likelihood = -258.93288  
Iteration 6: log likelihood = -258.93266  
Iteration 7: log likelihood = -258.93262  
Iteration 8: log likelihood = -258.93262

Multinomial logistic regression	Number of obs	=	278
	LR chi2(30)	=	59.05
	Prob > chi2	=	0.0012
Log likelihood = -258.93262	Pseudo R2	=	0.1024

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.040236	.0344054	1.19	0.233	.9749419	1.109903
	exper	.9276889	.0330918	-2.10	0.035	.8650456	.9948686
	male	1.35019	.6608073	0.61	0.540	.5173701	3.523613
	age	1.003614	.0268543	0.13	0.893	.9523372	1.057652
pyacctrati~2		.9419945	.3115964	-0.18	0.857	.4925891	1.801407
pyperstudlep		.9826276	.0261652	-0.66	0.510	.9326599	1.035272
pyperstude~s		1.031145	.0162809	1.94	0.052	.9997239	1.063554
pyperstudw~e		.9051361	.20323	-0.44	0.657	.5828996	1.40551
pyperstudb~k		.8863631	.2180535	-0.49	0.624	.5472784	1.435539
pyperstudl~o		.9024818	.2014164	-0.46	0.646	.5827292	1.397688
pyperstudm~e		.9965856	.0298171	-0.11	0.909	.9398257	1.056774
	Asian	(omitted)					
	Black	10.91673	14.16506	1.84	0.065	.8582467	138.8588
	Latino	3.812594	2.084959	2.45	0.014	1.305364	11.1355
	Other	2.53e+07	9.47e+10	0.00	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	.8784341	.7266994	-0.16	0.875	.1735955	4.445084
	OtherMale	(omitted)					
3							
	tenure	.996016	.0204971	-0.19	0.846	.9566419	1.037011
	exper	1.00515	.0250357	0.21	0.837	.9572599	1.055437
	male	1.204654	.4049459	0.55	0.580	.6233468	2.328063
	age	.9773385	.0217832	-1.03	0.304	.9355633	1.020979
pyacctrati~2		1.723468	.3838003	2.44	0.015	1.113906	2.6666
pyperstudlep		.981615	.0234321	-0.78	0.437	.9367467	1.028632
pyperstude~s		1.02928	.0119619	2.48	0.013	1.0061	1.052993
pyperstudw~e		1.219886	.2442812	0.99	0.321	.8238858	1.806225
pyperstudb~k		1.139719	.2486008	0.60	0.549	.7432413	1.747696
pyperstudl~o		1.212346	.2421581	0.96	0.335	.8196049	1.793281
pyperstudm~e		1.001519	.0285066	0.05	0.957	.9471766	1.058978
	Asian	(omitted)					
	Black	1.20e-06	.0015594	-0.01	0.992	0	.
	Latino	.8509106	.4543483	-0.30	0.762	.2987998	2.423191
	Other	.442259	2481.418	-0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	(omitted)					
	LatinoMale	1.037718	.8596601	0.04	0.964	.2046125	5.262918
	OtherMale	(omitted)					

-----> region = 16

note: Asian omitted because of collinearity  
 note: AsianMale omitted because of collinearity  
 Iteration 0: log likelihood = -527.46226  
 Iteration 1: log likelihood = -479.49592  
 Iteration 2: log likelihood = -478.07336  
 Iteration 3: log likelihood = -478.00469  
 Iteration 4: log likelihood = -477.989  
 Iteration 5: log likelihood = -477.98554  
 Iteration 6: log likelihood = -477.98477  
 Iteration 7: log likelihood = -477.98465  
 Iteration 8: log likelihood = -477.98463

Multinomial logistic regression	Number of obs	=	483
	LR chi2(34)	=	98.96
	Prob > chi2	=	0.0000
Log likelihood = -477.98463	Pseudo R2	=	0.0938

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.057022	.0206753	2.84	0.005	1.017266	1.098331
	exper	.9405088	.0245824	-2.35	0.019	.8935415	.9899449
	male	.9016129	.240968	-0.39	0.698	.5339802	1.522352
	age	.9823784	.0208812	-0.84	0.403	.9422928	1.024169
pyacctrati~2		1.271998	.2418467	1.27	0.206	.8762871	1.846403
pyperstudlep		1.004485	.0160139	0.28	0.779	.9735834	1.036367
pyperstude~s		.9906203	.0097042	-0.96	0.336	.9717818	1.009824
pyperstudw~e		.9453261	.03747	-1.42	0.156	.8746664	1.021694
pyperstudb~k		.9805625	.0437446	-0.44	0.660	.8984661	1.07016
pyperstudl~o		.9531824	.0356853	-1.28	0.200	.8857449	1.025754
pyperstudm~e		.9886841	.0165012	-0.68	0.495	.9568657	1.021561
	Asian	(omitted)					
	Black	.917815	.7893434	-0.10	0.921	.1700987	4.952328
	Latino	5.061503	4.138785	1.98	0.047	1.019185	25.13657
	Other	1.773656	2.29505	0.44	0.658	.1404249	22.40241
	AsianMale	(omitted)					
	BlackMale	326557.4	2.34e+08	0.02	0.986	0	.
	LatinoMale	.1672586	.1799288	-1.66	0.096	.0203098	1.377432
	OtherMale	9413962	1.76e+10	0.01	0.993	0	.
3							
	tenure	1.029929	.0165665	1.83	0.067	.9979656	1.062916
	exper	.9959814	.0220389	-0.18	0.856	.9537093	1.040127
	male	2.182289	.5628041	3.03	0.002	1.316408	3.617712
	age	1.001794	.0204786	0.09	0.930	.9624496	1.042746
pyacctrati~2		1.858932	.3318252	3.47	0.001	1.310151	2.63758
pyperstudlep		1.043748	.0161831	2.76	0.006	1.012506	1.075953
pyperstude~s		1.003459	.0089216	0.39	0.698	.9861248	1.021099
pyperstudw~e		1.020591	.0461914	0.45	0.652	.9339572	1.115262
pyperstudb~k		1.030064	.0520572	0.59	0.558	.9329246	1.137319
pyperstudl~o		1.004116	.0436191	0.09	0.925	.9221624	1.093353
pyperstudm~e		.9639379	.0229155	-1.54	0.122	.9200545	1.009914
	Asian	(omitted)					
	Black	1.286312	1.282056	0.25	0.801	.1823733	9.07259
	Latino	6.71126	5.73612	2.23	0.026	1.256844	35.8366
	Other	1.234899	1.814135	0.14	0.886	.0693697	21.98331
	AsianMale	(omitted)					
	BlackMale	136937.8	9.81e+07	0.02	0.987	0	.
	LatinoMale	.1596307	.1722902	-1.70	0.089	.0192491	1.323799
	OtherMale	.578713	1591.884	-0.00	1.000	0	.

-----> region = 17

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -513.40171  
 Iteration 1: log likelihood = -458.6916  
 Iteration 2: log likelihood = -455.62845  
 Iteration 3: log likelihood = -455.28295  
 Iteration 4: log likelihood = -455.22573  
 Iteration 5: log likelihood = -455.2128  
 Iteration 6: log likelihood = -455.20955  
 Iteration 7: log likelihood = -455.20893  
 Iteration 8: log likelihood = -455.20883  
 Iteration 9: log likelihood = -455.20881

Multinomial logistic regression	Number of obs	=	511
	LR chi2(36)	=	116.39
	Prob > chi2	=	0.0000
Log likelihood = -455.20881	Pseudo R2	=	0.1133

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.014452	.0222762	0.65	0.513	.9717173	1.059065
	exper	.9973685	.0287461	-0.09	0.927	.9425889	1.055332
	male	1.589629	.4676178	1.58	0.115	.8930991	2.829384
	age	.9853919	.023884	-0.61	0.544	.9396745	1.033333
pyacctrati~2		.7486588	.157203	-1.38	0.168	.496077	1.129845
pyperstudlep		1.030756	.0297162	1.05	0.293	.9741284	1.090676
pyperstude~s		.9899423	.0113048	-0.89	0.376	.9680315	1.012349
pyperstudw~e		1.066067	.0896944	0.76	0.447	.9039993	1.257191
pyperstudb~k		1.072398	.0922181	0.81	0.416	.9060647	1.269267
pyperstudl~o		1.053966	.0884088	0.63	0.531	.8941819	1.242301
pyperstudm~e		1.010845	.0131069	0.83	0.405	.9854802	1.036864
	Asian	4.38e-07	.0011857	-0.01	0.996	0	.
	Black	6.50e-07	.000743	-0.01	0.990	0	.
	Latino	.7550292	.5150505	-0.41	0.680	.1982935	2.874876
	Other	5.97e-07	.001614	-0.01	0.996	0	.
	AsianMale	(omitted)					
	BlackMale	2.83e+12	3.72e+15	0.02	0.983	0	.
	LatinoMale	5.060202	4.310158	1.90	0.057	.9530922	26.86587
	OtherMale	1.52132	4784.356	0.00	1.000	0	.
3							
	tenure	1.042821	.0174737	2.50	0.012	1.009129	1.077637
	exper	1.001446	.0218202	0.07	0.947	.959579	1.045139
	male	2.965432	.7365042	4.38	0.000	1.822556	4.824976
	age	1.021081	.019514	1.09	0.275	.9835415	1.060053
pyacctrati~2		1.500819	.2627079	2.32	0.020	1.064956	2.115072
pyperstudlep		1.02295	.0215564	1.08	0.282	.9815613	1.066085
pyperstude~s		1.029442	.0107646	2.77	0.006	1.008559	1.050758
pyperstudw~e		1.060872	.0765128	0.82	0.413	.9210262	1.221951
pyperstudb~k		1.033577	.0766102	0.45	0.656	.8938214	1.195185
pyperstudl~o		1.04883	.0757055	0.66	0.509	.9104681	1.208219
pyperstudm~e		.9780046	.0179616	-1.21	0.226	.9434267	1.01385
	Asian	7.05e-07	.0013777	-0.01	0.994	0	.
	Black	1.951918	2.514181	0.52	0.604	.1563384	24.37011
	Latino	.8005759	.4152541	-0.43	0.668	.2896604	2.212666
	Other	1.02e-06	.0019988	-0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	1582078	1.03e+09	0.02	0.982	0	.
	LatinoMale	1.810616	1.385878	0.78	0.438	.4039256	8.116171
	OtherMale	1246626	2.43e+09	0.01	0.994	0	.

-----> region = 18

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -429.12493  
 Iteration 1: log likelihood = -386.92855  
 Iteration 2: log likelihood = -384.20424  
 Iteration 3: log likelihood = -384.13484  
 Iteration 4: log likelihood = -384.11988  
 Iteration 5: log likelihood = -384.11616  
 Iteration 6: log likelihood = -384.11545  
 Iteration 7: log likelihood = -384.11531  
 Iteration 8: log likelihood = -384.1153

Multinomial logistic regression	Number of obs	=	404
	LR chi2(36)	=	90.02
	Prob > chi2	=	0.0000
Log likelihood = -384.1153	Pseudo R2	=	0.1049

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.017109	.0230657	0.75	0.454	.9728914 1.063337
exper	.9631581	.0253881	-1.42	0.154	.914662 1.014226
male	.7045029	.2508972	-0.98	0.325	.350543 1.415873
age	1.001965	.0204042	0.10	0.923	.9627611 1.042765
pyacctrati~2	.6568723	.1721501	-1.60	0.109	.3930086 1.097893
pyperstudlep	.9794514	.0203671	-1.00	0.318	.9403352 1.020195
pyperstude~s	.9851236	.0110311	-1.34	0.181	.9637386 1.006983
pyperstudw~e	1.023484	.1252911	0.19	0.850	.8051563 1.301014
pyperstudb~k	1.116875	.1442713	0.86	0.392	.8670641 1.438658
pyperstudl~o	1.043753	.1260261	0.35	0.723	.8237988 1.322436
pyperstudm~e	1.015321	.0155188	0.99	0.320	.9853558 1.046197
Asian	3.52e-06	.0049019	-0.01	0.993	0 .
Black	2.946187	2.675782	1.19	0.234	.4968015 17.4718
Latino	2.116843	.8409254	1.89	0.059	.9717285 4.611394
Other	.6689498	1789.444	-0.00	1.000	0 .
AsianMale	(omitted)				
BlackMale	.4345843	.6820938	-0.53	0.595	.0200487 9.420234
LatinoMale	.9913051	.5726132	-0.02	0.988	.3195383 3.07533
OtherMale	2.02928	7676.824	0.00	1.000	0 .
3					
tenure	1.007155	.018158	0.40	0.693	.9721873 1.04338
exper	1.046261	.0264537	1.79	0.074	.9956764 1.099416
male	1.115181	.3538145	0.34	0.731	.5988033 2.076858
age	.9835917	.0211741	-0.77	0.442	.9429545 1.02598
pyacctrati~2	1.447453	.340168	1.57	0.116	.9131924 2.294281
pyperstudlep	.965008	.0164575	-2.09	0.037	.933285 .9978093
pyperstude~s	1.03363	.0115894	2.95	0.003	1.011163 1.056596
pyperstudw~e	1.126592	.1257467	1.07	0.286	.9052281 1.402087
pyperstudb~k	1.149445	.1366806	1.17	0.241	.9104823 1.451124
pyperstudl~o	1.122497	.1253183	1.04	0.301	.9018931 1.397061
pyperstudm~e	.9293071	.0191633	-3.56	0.000	.8924965 .9676359
Asian	7.12e-07	.0009136	-0.01	0.991	0 .
Black	3.249937	3.470126	1.10	0.270	.4008697 26.34795
Latino	1.458475	.5680777	0.97	0.333	.6797594 3.12927
Other	1.13e+07	1.79e+10	0.01	0.992	0 .
AsianMale	(omitted)				
BlackMale	1.156223	1.661213	0.10	0.920	.0691945 19.32022
LatinoMale	.604222	.3498336	-0.87	0.384	.1942524 1.879433
OtherMale	.4150703	928.4922	-0.00	1.000	0 .

-----> region = 19

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -891.50929  
 Iteration 1: log likelihood = -821.81299  
 Iteration 2: log likelihood = -818.76136  
 Iteration 3: log likelihood = -818.47253  
 Iteration 4: log likelihood = -818.4276  
 Iteration 5: log likelihood = -818.41735  
 Iteration 6: log likelihood = -818.41481  
 Iteration 7: log likelihood = -818.41432  
 Iteration 8: log likelihood = -818.41424  
 Iteration 9: log likelihood = -818.41422

Multinomial logistic regression	Number of obs	=	827
	LR chi2(36)	=	146.19
	Prob > chi2	=	0.0000
Log likelihood = -818.41422	Pseudo R2	=	0.0820

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	.99894	.013993	-0.08	0.940	.9718873	1.026746
exper	1.040283	.0179586	2.29	0.022	1.005674	1.076084
male	.9992657	.3067604	-0.00	0.998	.5474874	1.823844
age	.9987879	.0126571	-0.10	0.924	.9742861	1.023906
pyacctrati~2	.9521497	.1565016	-0.30	0.765	.6899159	1.314057
pyperstudlep	.9929135	.0062007	-1.14	0.255	.9808344	1.005141
pyperstude~s	.9870928	.0099791	-1.29	0.199	.9677267	1.006846
pyperstudw~e	.887784	.0940153	-1.12	0.261	.7213832	1.092568
pyperstudb~k	.9799835	.1029208	-0.19	0.847	.7976696	1.203967
pyperstudl~o	.9403606	.0904956	-0.64	0.523	.7787159	1.135559
pyperstudm~e	1.01848	.0158662	1.18	0.240	.9878527	1.050057
Asian	1.343886	1.945893	0.20	0.838	.0786786	22.95451
Black	.9015496	.4290747	-0.22	0.828	.354713	2.291407
Latino	1.46675	.3669856	1.53	0.126	.8982202	2.395132
Other	.325446	.3746248	-0.98	0.329	.0340919	3.106756
AsianMale	(omitted)					
BlackMale	1.490521	1.319103	0.45	0.652	.2630449	8.445911
LatinoMale	.7698462	.2849485	-0.71	0.480	.3726858	1.590249
OtherMale	2.17e-06	.0018656	-0.02	0.988	0	.
3						
tenure	.9905053	.0149533	-0.63	0.527	.9616268	1.020251
exper	1.135564	.0243796	5.92	0.000	1.088772	1.184367
male	.6590083	.2461652	-1.12	0.264	.3169086	1.370401
age	.9742815	.0167933	-1.51	0.131	.941917	1.007758
pyacctrati~2	1.600409	.2868003	2.62	0.009	1.126398	2.273893
pyperstudlep	1.012239	.0070898	1.74	0.082	.9984378	1.02623
pyperstude~s	1.010443	.011975	0.88	0.381	.9872425	1.034188
pyperstudw~e	.9282262	.1061985	-0.65	0.515	.7417673	1.161555
pyperstudb~k	.9231059	.1060515	-0.70	0.486	.7369887	1.156225
pyperstudl~o	.9141924	.0952537	-0.86	0.389	.7453276	1.121316
pyperstudm~e	1.02299	.0183552	1.27	0.205	.9876397	1.059606
Asian	1.52e-06	.0020469	-0.01	0.992	0	.
Black	.3565463	.2528791	-1.45	0.146	.0887983	1.431619
Latino	.9422665	.2648923	-0.21	0.832	.5431026	1.634804
Other	5.00e-07	.0004068	-0.02	0.986	0	.
AsianMale	(omitted)					
BlackMale	2.75569	3.765742	0.74	0.458	.1892497	40.12597
LatinoMale	2.083474	.9235763	1.66	0.098	.8739	4.967231

OtherMale		6609796	5.37e+09	0.02	0.985	0	.
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-----> region = 20

```
Iteration 0: log likelihood = -2195.347
Iteration 1: log likelihood = -2097.3114
Iteration 2: log likelihood = -2094.2368
Iteration 3: log likelihood = -2094.1391
Iteration 4: log likelihood = -2094.1173
Iteration 5: log likelihood = -2094.1121
Iteration 6: log likelihood = -2094.1111
Iteration 7: log likelihood = -2094.1109
Iteration 8: log likelihood = -2094.1108
```

Multinomial logistic regression	Number of obs	=	2112
	LR chi2(38)	=	202.47
	Prob > chi2	=	0.0000
Log likelihood = -2094.1108	Pseudo R2	=	0.0461

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)					
2						
	tenure	.9932013	.0079503	-0.85	0.394	.9777405 1.008907
	exper	1.00777	.0103276	0.76	0.450	.9877302 1.028216
	male	1.066399	.1649887	0.42	0.678	.7874545 1.444156
	age	.9853498	.0083251	-1.75	0.081	.9691672 1.001803
	pyacctrati~2	.9728388	.0882124	-0.30	0.761	.8144379 1.162047
	pyperstudlep	.987079	.0066345	-1.93	0.053	.9741608 1.000168
	pyperstude~s	1.00442	.0055167	0.80	0.422	.9936656 1.015291
	pyperstudw~e	1.001014	.029894	0.03	0.973	.9441043 1.061353
	pyperstudb~k	.9894981	.032847	-0.32	0.750	.9271688 1.056017
	pyperstudl~o	.995141	.0308298	-0.16	0.875	.9365137 1.057438
	pyperstudm~e	1.003136	.0077778	0.40	0.686	.9880076 1.018497
	Asian	.6935113	.8063704	-0.31	0.753	.0710119 6.77292
	Black	1.487098	.4385331	1.35	0.178	.8343091 2.650647
	Latino	1.196493	.1719941	1.25	0.212	.9027159 1.585875
	Other	.6852935	.4702639	-0.55	0.582	.1785512 2.63021
	AsianMale	1545701	8.37e+08	0.03	0.979	0 .
	BlackMale	1.651078	.7571103	1.09	0.274	.672117 4.05593
	LatinoMale	1.464339	.3318792	1.68	0.092	.9391272 2.283278
	OtherMale	.4361207	.5615104	-0.64	0.519	.0349683 5.439244
3						
	tenure	1.009312	.0082254	1.14	0.255	.9933184 1.025563
	exper	1.076989	.0137614	5.80	0.000	1.050352 1.104301
	male	1.285961	.2207906	1.46	0.143	.9185067 1.800417
	age	.9616305	.0107782	-3.49	0.000	.9407359 .9829891
	pyacctrati~2	1.275556	.1255366	2.47	0.013	1.051784 1.546936
	pyperstudlep	1.012546	.0062073	2.03	0.042	1.000453 1.024786
	pyperstude~s	1.027896	.0063581	4.45	0.000	1.015509 1.040433
	pyperstudw~e	1.004027	.0324644	0.12	0.901	.9423726 1.069716
	pyperstudb~k	.9820763	.0350173	-0.51	0.612	.9157869 1.053164
	pyperstudl~o	.9797151	.0328075	-0.61	0.541	.9174783 1.046174
	pyperstudm~e	.9956421	.0093271	-0.47	0.641	.9775282 1.014092
	Asian	1.299841	1.21511	0.28	0.779	.2080535 8.120923
	Black	1.613585	.4981572	1.55	0.121	.881055 2.955158
	Latino	.7974922	.1307546	-1.38	0.168	.5783168 1.099732
	Other	.3752089	.4027538	-0.91	0.361	.0457702 3.075837
	AsianMale	519298.8	2.81e+08	0.02	0.981	0 .
	BlackMale	1.03421	.5490276	0.06	0.949	.3653711 2.927408
	LatinoMale	1.21408	.31924	0.74	0.461	.7251458 2.03268
	OtherMale	1.016319	1.569728	0.01	0.992	.0492414 20.97635

-----

by pyschooltype:

```
. bysort pyschooltype : mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2009a==1, rrr iter(20)
```

-----> pyschooltype =  
no observations

-----> pyschooltype = B

```
note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -388.2121
Iteration 1: log likelihood = -350.51408
Iteration 2: log likelihood = -344.5998
Iteration 3: log likelihood = -340.71804
Iteration 4: log likelihood = -339.56129
Iteration 5: log likelihood = -338.38926
Iteration 6: log likelihood = -338.38383
Iteration 7: log likelihood = -338.3831
Iteration 8: log likelihood = -338.38296
Iteration 9: log likelihood = -338.38293
Iteration 10: log likelihood = -338.38293
```

```
Multinomial logistic regression      Number of obs   =      426
                                     LR chi2(34)       =      99.66
                                     Prob > chi2      =      0.0000
Log likelihood = -338.38293          Pseudo R2       =      0.1284
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.019621	.0283824	0.70	0.485	.9654828	1.076795
	exper	1.006709	.0375499	0.18	0.858	.9357388	1.083063
	male	1.028254	.4609402	0.06	0.950	.4270989	2.475555
	age	.9504232	.0300136	-1.61	0.107	.8933811	1.011107
	pyacctrati~2	.7557543	.2194753	-0.96	0.335	.4277469	1.335286
	pyperstudlep	1.02388	.0217508	1.11	0.267	.982125	1.067411
	pyperstude~s	.9762229	.0138872	-1.69	0.091	.9493804	1.003824
	pyperstudw~e	.9804513	.0294589	-0.66	0.511	.9243802	1.039924
	pyperstudb~k	1.026689	.0345585	0.78	0.434	.9611415	1.096707
	pyperstudl~o	1.010766	.0329551	0.33	0.743	.9481961	1.077466
	pyperstudm~e	1.004076	.0130622	0.31	0.755	.9787984	1.030007
	Asian	.7906221	2742.798	-0.00	1.000	0	.
	Black	1.673662	1.267592	0.68	0.497	.3793043	7.384952
	Latino	1.20513	1.118855	0.20	0.841	.1953306	7.435288
	Other	6.545215	22706.42	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	2.591456	2.943339	0.84	0.402	.279744	24.00638
	LatinoMale	3.330061	3.842982	1.04	0.297	.3468504	31.97143
	OtherMale	(omitted)					
3							
	tenure	.9703576	.0173884	-1.68	0.093	.9368685	1.005044
	exper	.9738492	.0181341	-1.42	0.155	.9389477	1.010048
	male	1.255472	.3203243	0.89	0.373	.7614285	2.070068
	age	1.026835	.0173262	1.57	0.117	.9934316	1.061361
	pyacctrati~2	.7917743	.1466015	-1.26	0.207	.5508034	1.138167

pyperstudlep		.983783	.0178835	-0.90	0.368	.949349	1.019466
pyperstude~s		1.015339	.0096565	1.60	0.109	.9965877	1.034443
pyperstudw~e		1.010073	.0300925	0.34	0.737	.9527822	1.07081
pyperstudb~k		.9969848	.032601	-0.09	0.926	.9350925	1.062974
pyperstudl~o		1.010058	.0309336	0.33	0.744	.9512131	1.072543
pyperstudm~e		.9566902	.0173347	-2.44	0.015	.9233111	.991276
Asian		1392976	1.91e+09	0.01	0.992	0	.
Black		.4709076	.5387946	-0.66	0.510	.0500061	4.434535
Latino		.8083231	.7090089	-0.24	0.808	.1448658	4.510286
Other		933523.5	1.28e+09	0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		9.464802	12.84566	1.66	0.098	.6620005	135.3209
LatinoMale		4.536561	4.803206	1.43	0.153	.5695008	36.13759
OtherMale		(omitted)					

-----> pyschooltype = E

Iteration 0: log likelihood = -12214.373  
Iteration 1: log likelihood = -11582.932  
Iteration 2: log likelihood = -11577.177  
Iteration 3: log likelihood = -11577.173  
Iteration 4: log likelihood = -11577.173

Multinomial logistic regression	Number of obs	=	11232
	LR chi2(38)	=	1274.40
	Prob > chi2	=	0.0000
Log likelihood = -11577.173	Pseudo R2	=	0.0522

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
	tenure	1.014929	.0041721	3.60	0.000	1.006785 1.023139
	exper	1.007191	.0051486	1.40	0.161	.9971504 1.017333
	male	1.650231	.1403344	5.89	0.000	1.39688 1.949531
	age	.9845499	.0037993	-4.03	0.000	.9771314 .9920246
pyacctrati~2		.9005668	.0334577	-2.82	0.005	.8373215 .9685893
pyperstudlep		1.00593	.0017161	3.47	0.001	1.002572 1.009299
pyperstude~s		.9918013	.0020723	-3.94	0.000	.987748 .9958712
pyperstudw~e		.9972148	.004889	-0.57	0.569	.9876783 1.006843
pyperstudb~k		.9953096	.005631	-0.83	0.406	.984334 1.006408
pyperstudl~o		.9986714	.0051874	-0.26	0.798	.9885559 1.00889
pyperstudm~e		1.006732	.0037691	1.79	0.073	.9993716 1.014147
Asian		.6163851	.2350086	-1.27	0.204	.2919546 1.301335
Black		.9600183	.0874684	-0.45	0.654	.8030185 1.147713
Latino		1.129946	.0821768	1.68	0.093	.9798345 1.303054
Other		1.719171	.4514358	2.06	0.039	1.02755 2.876308
AsianMale		1.478484	1.100146	0.53	0.599	.3439016 6.356221
BlackMale		1.156593	.22986	0.73	0.464	.7834543 1.707448
LatinoMale		.6361956	.0889704	-3.23	0.001	.4836733 .8368144
OtherMale		.3388777	.2324761	-1.58	0.115	.0883291 1.300116
3						
	tenure	1.019577	.0036489	5.42	0.000	1.01245 1.026753
	exper	1.067441	.0052392	13.30	0.000	1.057222 1.077759
	male	2.436892	.1918985	11.31	0.000	2.088366 2.843584
	age	.9889676	.0039478	-2.78	0.005	.9812602 .9967355
pyacctrati~2		.9338645	.0344291	-1.86	0.063	.8687649 1.003842
pyperstudlep		.9978068	.0016845	-1.30	0.193	.9945108 1.001114
pyperstude~s		1.005796	.0020258	2.87	0.004	1.001833 1.009774
pyperstudw~e		.9990338	.0049428	-0.20	0.845	.9893929 1.008769
pyperstudb~k		.9868013	.0056207	-2.33	0.020	.9758461 .9978794
pyperstudl~o		.9916723	.0052005	-1.59	0.111	.9815317 1.001918



pyperstudm~e		.9979339	.0038429	-0.54	0.591	.9904302	1.005494
Asian		1.028433	.3728122	0.08	0.938	.505372	2.092864
Black		.9100143	.084273	-1.02	0.309	.7589648	1.091126
Latino		1.040423	.0768069	0.54	0.591	.9002678	1.202397
Other		1.260086	.3749453	0.78	0.437	.7032676	2.25777
AsianMale		.93157	.6743865	-0.10	0.922	.2254289	3.849651
BlackMale		.7408043	.151796	-1.46	0.143	.495777	1.106931
LatinoMale		.5477641	.0748909	-4.40	0.000	.4190023	.7160951
OtherMale		.5304616	.3471271	-0.97	0.333	.1471089	1.912798

-----> pyschooltype = M

Iteration 0: log likelihood = -6653.1301  
Iteration 1: log likelihood = -6351.1693  
Iteration 2: log likelihood = -6343.4012  
Iteration 3: log likelihood = -6343.3732  
Iteration 4: log likelihood = -6343.3732

Multinomial logistic regression	Number of obs	=	6328
	LR chi2(38)	=	619.51
	Prob > chi2	=	0.0000
Log likelihood = -6343.3732	Pseudo R2	=	0.0466

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.006389	.0048969	1.31	0.191	.9968363 1.016032
exper		1.010316	.0058797	1.76	0.078	.9988571 1.021906
male		1.42539	.1083869	4.66	0.000	1.228028 1.654471
age		.9810601	.0044927	-4.18	0.000	.972294 .9899053
pyacctrati~2		.8960229	.0488177	-2.02	0.044	.8052735 .9969992
pyperstudlep		1.005463	.003472	1.58	0.115	.998681 1.012291
pyperstude~s		.9971942	.0027461	-1.02	0.308	.9918265 1.002591
pyperstudw~e		.9897835	.0066366	-1.53	0.126	.9768611 1.002877
pyperstudb~k		.9902353	.0076168	-1.28	0.202	.9754187 1.005277
pyperstudl~o		.9907754	.0072208	-1.27	0.204	.9767234 1.00503
pyperstudm~e		1.0087	.004663	1.87	0.061	.9996019 1.017881
Asian		.7684503	.3330094	-0.61	0.543	.3286593 1.796742
Black		1.098919	.1164164	0.89	0.373	.8928774 1.352507
Latino		1.628005	.1818767	4.36	0.000	1.307861 2.026515
Other		.6777485	.2342956	-1.13	0.261	.3441999 1.334524
AsianMale		2.365014	1.620859	1.26	0.209	.6172541 9.061573
BlackMale		1.334208	.2226508	1.73	0.084	.9620028 1.850423
LatinoMale		1.121493	.1729982	0.74	0.457	.8288816 1.517401
OtherMale		2.022895	1.054559	1.35	0.177	.7281687 5.619719
3						
tenure		1.00903	.0051562	1.76	0.079	.9989745 1.019187
exper		1.087107	.0080688	11.25	0.000	1.071407 1.103037
male		2.101466	.1809032	8.63	0.000	1.7752 2.487697
age		.9527239	.0061747	-7.47	0.000	.9406982 .9649034
pyacctrati~2		1.036024	.0661924	0.55	0.580	.9140837 1.174232
pyperstudlep		.9890593	.0046037	-2.36	0.018	.9800772 .9981237
pyperstude~s		1.013345	.0033066	4.06	0.000	1.006885 1.019847
pyperstudw~e		1.011074	.0091504	1.22	0.224	.9932981 1.029169
pyperstudb~k		.9955154	.010287	-0.43	0.664	.975556 1.015883
pyperstudl~o		.9981279	.0097347	-0.19	0.848	.9792295 1.017391
pyperstudm~e		1.003002	.0058114	0.52	0.605	.9916766 1.014457
Asian		.8705448	.4941354	-0.24	0.807	.2861774 2.648177
Black		.8220705	.1216414	-1.32	0.185	.6151162 1.098654
Latino		1.461078	.2083456	2.66	0.008	1.104827 1.932201
Other		.8810271	.3811661	-0.29	0.770	.3773348 2.057083

AsianMale		.7724163	.7731664	-0.26	0.796	.1085984	5.493885
BlackMale		1.506797	.3187299	1.94	0.053	.9954093	2.280908
LatinoMale		1.05734	.1957357	0.30	0.763	.7355961	1.519812
OtherMale		1.093873	.6956587	0.14	0.888	.3145121	3.80449

-----> pyschooltype = S

Iteration 0: log likelihood = -7454.0579  
Iteration 1: log likelihood = -7164.8277  
Iteration 2: log likelihood = -7156.2983  
Iteration 3: log likelihood = -7156.2504  
Iteration 4: log likelihood = -7156.2504

Multinomial logistic regression	Number of obs	=	7771
	LR chi2(38)	=	595.61
	Prob > chi2	=	0.0000
Log likelihood = -7156.2504	Pseudo R2	=	0.0400

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.007531	.0037738	2.00	0.045	1.000162 1.014955
exper		1.008618	.0047476	1.82	0.068	.9993553 1.017966
male		1.078635	.0687885	1.19	0.235	.9518981 1.222247
age		.9760384	.0039112	-6.05	0.000	.9684027 .9837344
pyacctrati~2		.8748577	.0531139	-2.20	0.028	.7767116 .9854058
pypersstudlep		1.023382	.0045062	5.25	0.000	1.014588 1.032252
pypersstude~s		.9899968	.0021962	-4.53	0.000	.9857017 .9943106
pypersstudw~e		.9945133	.0060561	-0.90	0.366	.9827141 1.006454
pypersstudb~k		.9966903	.0068225	-0.48	0.628	.9834077 1.010152
pypersstudl~o		.9948891	.0064577	-0.79	0.430	.9823125 1.007627
pypersstudm~e		1.003528	.0034299	1.03	0.303	.996828 1.010273
Asian		1.315584	.5554916	0.65	0.516	.5750523 3.009744
Black		1.423309	.148681	3.38	0.001	1.159795 1.746695
Latino		1.917507	.2021353	6.18	0.000	1.559578 2.357583
Other		.9790021	.3381076	-0.06	0.951	.4975229 1.926434
AsianMale		.6784899	.4799702	-0.55	0.583	.1695881 2.71451
BlackMale		1.849547	.2748444	4.14	0.000	1.382216 2.474883
LatinoMale		1.141525	.1533203	0.99	0.324	.8773219 1.485292
OtherMale		.8577859	.433475	-0.30	0.761	.3185894 2.309545
3						
tenure		.998723	.0047635	-0.27	0.789	.9894301 1.008103
exper		1.057142	.0074626	7.87	0.000	1.042616 1.07187
male		1.995606	.1843932	7.48	0.000	1.665038 2.391803
age		.9684398	.0063111	-4.92	0.000	.9561489 .9808887
pyacctrati~2		1.128234	.0925206	1.47	0.141	.9607195 1.324957
pypersstudlep		.9895841	.0073392	-1.41	0.158	.9753035 1.004074
pypersstude~s		1.012709	.0030475	4.20	0.000	1.006753 1.018699
pypersstudw~e		1.026795	.0110274	2.46	0.014	1.005408 1.048638
pypersstudb~k		1.015335	.012036	1.28	0.199	.9920167 1.039201
pypersstudl~o		1.015156	.0114225	1.34	0.181	.9930139 1.037793
pypersstudm~e		.9928796	.0056998	-1.24	0.213	.9817709 1.004114
Asian		.5990517	.6255671	-0.49	0.624	.0773725 4.63812
Black		.9440826	.1912767	-0.28	0.776	.6346752 1.404328
Latino		1.619101	.2790015	2.80	0.005	1.155037 2.269614
Other		.6271933	.4641152	-0.63	0.528	.1470696 2.67473
AsianMale		.9503194	1.416015	-0.03	0.973	.0512324 17.62765
BlackMale		1.53331	.391889	1.67	0.094	.9291306 2.530365
LatinoMale		.7848308	.1617487	-1.18	0.240	.5240211 1.175448
OtherMale		2.487871	2.076035	1.09	0.275	.4847661 12.76802

## 2009-10

```
. mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstuddecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
> Male LatinoMale OtherMale if validcert2010a==1, rrr iter(20)
```

```
Iteration 0: log likelihood = -28524.928
Iteration 1: log likelihood = -27223.492
Iteration 2: log likelihood = -27193.717
Iteration 3: log likelihood = -27193.655
Iteration 4: log likelihood = -27193.655
```

```
Multinomial logistic regression      Number of obs   =      27019
                                     LR chi2(38)       =      2662.55
                                     Prob > chi2       =      0.0000
Log likelihood = -27193.655          Pseudo R2       =      0.0467
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.0115	.0024003	4.82	0.000	1.006806	1.016215
	exper	1.016564	.0029838	5.60	0.000	1.010733	1.022429
	male	1.173072	.0460167	4.07	0.000	1.086261	1.266821
	age	.9780635	.0022937	-9.46	0.000	.9735782	.9825693
pyacctrati~2		.9465964	.0185541	-2.80	0.005	.9109206	.9836693
pyperstudlep		1.00469	.0011471	4.10	0.000	1.002445	1.006941
pyperstude~s		.9932212	.0012139	-5.57	0.000	.9908449	.9956032
pyperstudw~e		.9930673	.0028905	-2.39	0.017	.9874183	.9987487
pyperstudb~k		.995728	.0033473	-1.27	0.203	.989189	1.00231
pyperstudl~o		.9974023	.003118	-0.83	0.405	.9913099	1.003532
pyperstudm~e		1.003337	.0021545	1.55	0.121	.9991228	1.007568
	Asian	.922873	.1989886	-0.37	0.710	.6047935	1.40824
	Black	1.137561	.062369	2.35	0.019	1.021659	1.266612
	Latino	1.371203	.0690047	6.27	0.000	1.242412	1.513344
	Other	.9028993	.1559034	-0.59	0.554	.6436689	1.266532
	AsianMale	.8219976	.3019733	-0.53	0.594	.4001002	1.688777
	BlackMale	1.598778	.1444119	5.19	0.000	1.339375	1.908421
	LatinoMale	1.164392	.0877757	2.02	0.043	1.004461	1.349789
	OtherMale	1.055324	.3046223	0.19	0.852	.5993545	1.858179
3							
	tenure	1.01484	.0023635	6.33	0.000	1.010219	1.019483
	exper	1.072785	.0035728	21.10	0.000	1.065806	1.079811
	male	1.402379	.0585598	8.10	0.000	1.292175	1.521982
	age	.9732214	.0027907	-9.47	0.000	.967767	.9787065
pyacctrati~2		1.358989	.029812	13.98	0.000	1.301797	1.418694
pyperstudlep		1.008145	.0012544	6.52	0.000	1.005689	1.010607
pyperstude~s		1.016274	.0013542	12.12	0.000	1.013623	1.018932
pyperstudw~e		1.010214	.0035422	2.90	0.004	1.003296	1.017181
pyperstudb~k		.9973405	.0040323	-0.66	0.510	.9894686	1.005275
pyperstudl~o		.9956069	.0037289	-1.18	0.240	.9883251	1.002942
pyperstudm~e		.9944904	.0025416	-2.16	0.031	.9895214	.9994843
	Asian	.7846258	.2118488	-0.90	0.369	.462209	1.331947
	Black	.8204675	.0551452	-2.94	0.003	.7192013	.9359923
	Latino	1.103263	.0633502	1.71	0.087	.9858307	1.234684
	Other	.8797684	.1753112	-0.64	0.520	.5953189	1.300131
	AsianMale	.791968	.3725111	-0.50	0.620	.315018	1.99104
	BlackMale	1.382321	.1557905	2.87	0.004	1.108349	1.724015
	LatinoMale	1.015896	.088318	0.18	0.856	.85674	1.204618
	OtherMale	1.162046	.3724128	0.47	0.639	.6200549	2.177793

```
. mlogtest, all
```

```
**** Likelihood-ratio tests for independent variables (N=27019)
```

```
Ho: All coefficients associated with given variable(s) are 0.
```

	chi2	df	P>chi2
-----+-----			
tenure	46.452	2	0.000
exper	480.511	2	0.000
male	67.351	2	0.000
age	142.528	2	0.000
pyacctrati~2	265.360	2	0.000
pyperstudlep	45.540	2	0.000
pyperstude~s	256.820	2	0.000
pyperstudw~e	21.921	2	0.000
pyperstudb~k	1.677	2	0.432
pyperstudl~o	1.584	2	0.453
pyperstudm~e	10.799	2	0.005
Asian	0.848	2	0.654
Black	20.722	2	0.000
Latino	39.533	2	0.000
Other	0.598	2	0.741
AsianMale	0.423	2	0.810
BlackMale	28.230	2	0.000
LatinoMale	4.526	2	0.104
OtherMale	0.221	2	0.895
-----+-----			

```
**** Wald tests for independent variables (N=27019)
```

```
Ho: All coefficients associated with given variable(s) are 0.
```

	chi2	df	P>chi2
-----+-----			
tenure	46.389	2	0.000
exper	446.114	2	0.000
male	67.430	2	0.000
age	139.854	2	0.000
pyacctrati~2	261.599	2	0.000
pyperstudlep	45.372	2	0.000
pyperstude~s	252.532	2	0.000
pyperstudw~e	21.538	2	0.000
pyperstudb~k	1.678	2	0.432
pyperstudl~o	1.586	2	0.452
pyperstudm~e	10.607	2	0.005
Asian	0.822	2	0.663
Black	20.511	2	0.000
Latino	39.706	2	0.000
Other	0.596	2	0.742
AsianMale	0.421	2	0.810
BlackMale	27.991	2	0.000
LatinoMale	4.530	2	0.104
OtherMale	0.221	2	0.895
-----+-----			

```
**** Hausman tests of IIA assumption (N=27019)
```

```
Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.
```

Omitted	chi2	df	P>chi2	evidence
-----+-----				
2	-79.051	20	---	---
3	3.864	19	1.000	for Ho
-----+-----				

```
Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.
```

\*\*\*\* Small-Hsiao tests of IIA assumption (N=27019)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted	lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2	-5578.609	-5571.443	14.333	20	0.813	for Ho
3	-6775.905	-6770.166	11.478	20	0.933	for Ho

\*\*\*\* Wald tests for combining alternatives (N=27019)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested	chi2	df	P>chi2
2- 3	1550.592	19	0.000
2- 1	431.759	19	0.000
3- 1	1797.253	19	0.000

\*\*\*\* LR tests for combining alternatives (N=27019)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested	chi2	df	P>chi2
2- 3	1711.204	19	0.000
2- 1	442.337	19	0.000
3- 1	2006.578	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstuddecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
> AsianMale BlackMale LatinoMale OtherMale if validcert2010a==1, rrr iter(20)
```

-----> region = 1

note: OtherMale omitted because of collinearity

```
Iteration 0: log likelihood = -2064.9562
Iteration 1: log likelihood = -1956.6008
Iteration 2: log likelihood = -1952.1061
Iteration 3: log likelihood = -1951.7462
Iteration 4: log likelihood = -1951.6823
Iteration 5: log likelihood = -1951.6761
Iteration 6: log likelihood = -1951.6747
Iteration 7: log likelihood = -1951.6744
Iteration 8: log likelihood = -1951.6744
```

```
Multinomial logistic regression      Number of obs   =      1926
                                      LR chi2(36)        =      226.56
                                      Prob > chi2       =      0.0000
Log likelihood = -1951.6744          Pseudo R2       =      0.0549
```

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

```

2      |
      tenure | 1.020178 .0088362 2.31 0.021 1.003006 1.037645
      exper | 1.003629 .0115474 0.31 0.753 .9812498 1.026519
      male | .6063288 .2024218 -1.50 0.134 .3151636 1.166488
      age | .9735791 .0094297 -2.76 0.006 .9552715 .9922376
pyaccrati~2 | .8927229 .0681759 -1.49 0.137 .7686199 1.036864
pyperstudlep | 1.001243 .0026955 0.46 0.644 .9959741 1.00654
pyperstude~s | .9993828 .0053023 -0.12 0.907 .9890443 1.009829
pyperstudw~e | .9315509 .0656245 -1.01 0.314 .8114139 1.069475
pyperstudb~k | .981266 .0745099 -0.25 0.803 .8455766 1.138729
pyperstudl~o | .9714419 .0564377 -0.50 0.618 .8668915 1.088602
pyperstudm~e | .9975814 .0084615 -0.29 0.775 .9811343 1.014304
      Asian | .6227006 .5586965 -0.53 0.598 .1072922 3.614018
      Black | .5348857 .636266 -0.53 0.599 .0519668 5.505494
      Latino | 1.137387 .245378 0.60 0.551 .745198 1.735982
      Other | .5361867 .4646108 -0.72 0.472 .0981165 2.930151
      AsianMale | 1.70e-06 .0029924 -0.01 0.994 0 .
      BlackMale | 9.263644 13.91493 1.48 0.138 .4877435 175.9431
      LatinoMale | 2.060234 .7294097 2.04 0.041 1.029332 4.123612
      OtherMale | (omitted)

```

```

3      |
      tenure | 1.028363 .0093094 3.09 0.002 1.010278 1.046772
      exper | 1.099836 .0173498 6.03 0.000 1.066352 1.134372
      male | .6009191 .2120819 -1.44 0.149 .3008861 1.200134
      age | .930622 .0134112 -4.99 0.000 .9047043 .9572821
pyaccrati~2 | 1.18758 .1055338 1.93 0.053 .997749 1.413529
pyperstudlep | 1.0154 .0031545 4.92 0.000 1.009236 1.021601
pyperstude~s | 1.000355 .0062495 0.06 0.955 .9881812 1.012679
pyperstudw~e | .9991923 .078672 -0.01 0.992 .8563065 1.16592
pyperstudb~k | 1.019559 .0974196 0.20 0.839 .8454333 1.229547
pyperstudl~o | .9970892 .0644246 -0.05 0.964 .8784874 1.131703
pyperstudm~e | .9858531 .0107972 -1.30 0.193 .9649164 1.007244
      Asian | 9.52e-07 .0006161 -0.02 0.983 0 .
      Black | 7.38e-07 .0006366 -0.02 0.987 0 .
      Latino | .6551021 .1460192 -1.90 0.058 .4232336 1.014
      Other | .1952572 .2236122 -1.43 0.154 .0206916 1.842549
      AsianMale | 12.53564 28733.08 0.00 0.999 0 .
      BlackMale | 7.880116 8650.721 0.00 0.998 0 .
      LatinoMale | 2.261391 .8610813 2.14 0.032 1.072159 4.76971
      OtherMale | (omitted)

```

-----> region = 2

```

Iteration 0: log likelihood = -753.73606
Iteration 1: log likelihood = -716.66905
Iteration 2: log likelihood = -715.40688
Iteration 3: log likelihood = -715.25866
Iteration 4: log likelihood = -715.23887
Iteration 5: log likelihood = -715.23647
Iteration 6: log likelihood = -715.23601
Iteration 7: log likelihood = -715.23591
Iteration 8: log likelihood = -715.23589

```

```

Multinomial logistic regression      Number of obs   =      730
                                      LR chi2(38)        =      77.00
                                      Prob > chi2         =      0.0002
Log likelihood = -715.23589          Pseudo R2        =      0.0511

```

```

-----+-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----
1          | (base outcome)
-----+-----
2          |

```

tenure		1.011474	.0158411	0.73	0.466	.9808976	1.043003
exper		.9965449	.0186272	-0.19	0.853	.960697	1.033731
male		1.246856	.352518	0.78	0.435	.7164075	2.170065
age		.986294	.0139697	-0.97	0.330	.9592904	1.014058
pyacctrati~2		1.012802	.1426651	0.09	0.928	.7684629	1.334832
pyperstudlep		1.021031	.0250932	0.85	0.397	.9730155	1.071417
pyperstude~s		1.002256	.0084179	0.27	0.788	.9858927	1.018892
pyperstudw~e		.9255787	.0672693	-1.06	0.287	.8026933	1.067277
pyperstudb~k		.9156989	.0793695	-1.02	0.310	.7726336	1.085255
pyperstudl~o		.9307904	.065684	-1.02	0.309	.8105584	1.068857
pyperstudm~e		.9860736	.0173304	-0.80	0.425	.9526851	1.020632
Asian		8428955	1.68e+10	0.01	0.994	0	.
Black		.9813996	.8832151	-0.02	0.983	.1681911	5.726492
Latino		.7851537	.2046531	-0.93	0.353	.4710706	1.30865
Other		1.949428	1.998758	0.65	0.515	.2613161	14.5428
AsianMale		1.16e-13	2.88e-10	-0.01	0.990	0	.
BlackMale		.6542348	.9621911	-0.29	0.773	.036632	11.68439
LatinoMale		1.524455	.6017706	1.07	0.285	.7032458	3.304623
OtherMale		3.90e-07	.0003394	-0.02	0.986	0	.

-----

3		tenure	1.011288	.0144211	0.79	0.431	.9834149	1.039952
		exper	1.023974	.0187769	1.29	0.196	.9878255	1.061445
		male	.6471191	.211205	-1.33	0.182	.3413276	1.226866
		age	1.004868	.0148666	0.33	0.743	.9761486	1.034433
		pyacctrati~2	1.431017	.1972141	2.60	0.009	1.092289	1.874789
		pyperstudlep	1.023779	.0236994	1.02	0.310	.9783669	1.071299
		pyperstude~s	1.006717	.0086043	0.78	0.433	.9899931	1.023723
		pyperstudw~e	1.096984	.083469	1.22	0.224	.9450021	1.273409
		pyperstudb~k	1.084514	.1004314	0.88	0.381	.9045025	1.300351
		pyperstudl~o	1.09765	.081464	1.26	0.209	.9490526	1.269513
		pyperstudm~e	.9442643	.0180047	-3.01	0.003	.9096271	.9802205
		Asian	1.694603	5843.894	0.00	1.000	0	.
		Black	2.989837	2.352501	1.39	0.164	.6395793	13.97657
		Latino	1.045884	.2619982	0.18	0.858	.6401097	1.708886
		Other	.8824054	1.107045	-0.10	0.921	.0754694	10.31728
		AsianMale	1.20e-06	.004495	-0.00	0.997	0	.
		BlackMale	.5493516	.7831979	-0.42	0.674	.0335979	8.982326
		LatinoMale	2.163431	.9253727	1.80	0.071	.935525	5.003003
		OtherMale	4.04e-06	.0032545	-0.02	0.988	0	.

-----

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -355.32467  
Iteration 1: log likelihood = -306.89744  
Iteration 2: log likelihood = -304.79014  
Iteration 3: log likelihood = -304.68829  
Iteration 4: log likelihood = -304.67598  
Iteration 5: log likelihood = -304.67371  
Iteration 6: log likelihood = -304.67348  
Iteration 7: log likelihood = -304.67343  
Iteration 8: log likelihood = -304.67342

Multinomial logistic regression	Number of obs	=	333
	LR chi2(32)	=	101.30
	Prob > chi2	=	0.0000
Log likelihood = -304.67342	Pseudo R2	=	0.1425

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

-----							
2							
tenure		1.021702	.0243096	0.90	0.367	.9751501	1.070477
exper		.9547492	.0271957	-1.63	0.104	.9029073	1.009568
male		1.324841	.4668175	0.80	0.425	.6641046	2.642962
age		.9940622	.0235175	-0.25	0.801	.9490211	1.041241
pyacctrati~2		.8190218	.171085	-0.96	0.339	.5438616	1.233396
pyperstudlep		1.025801	.0267749	0.98	0.329	.9746429	1.079644
pyperstude~s		.9879251	.0191564	-0.63	0.531	.9510838	1.026193
pyperstudw~e		.8947304	.0474935	-2.10	0.036	.8063235	.9928305
pyperstudb~k		.9707077	.053138	-0.54	0.587	.8719516	1.080649
pyperstudl~o		.926542	.0525701	-1.34	0.179	.8290289	1.035525
pyperstudm~e		.997858	.013707	-0.16	0.876	.9713511	1.025088
Asian		(omitted)					
Black		1.441684	1.01891	0.52	0.605	.360814	5.760455
Latino		.9945869	.6591522	-0.01	0.993	.2713424	3.64559
Other		2.196872	3.293405	0.52	0.600	.1163428	41.48297
AsianMale		(omitted)					
BlackMale		.7864826	1.267381	-0.15	0.882	.0334208	18.50809
LatinoMale		.3670222	.4102145	-0.90	0.370	.0410506	3.28145
OtherMale		(omitted)					

-----							
3							
tenure		1.009541	.0192034	0.50	0.618	.9725955	1.047889
exper		1.009279	.0250823	0.37	0.710	.961297	1.059657
male		1.902112	.643219	1.90	0.057	.9803775	3.690445
age		1.041216	.0231753	1.81	0.070	.9967692	1.087644
pyacctrati~2		1.309278	.2552031	1.38	0.167	.8935455	1.918436
pyperstudlep		1.01458	.0245865	0.60	0.550	.967518	1.063932
pyperstude~s		1.076161	.0193391	4.08	0.000	1.038917	1.11474
pyperstudw~e		1.078223	.0652317	1.24	0.213	.9576606	1.213964
pyperstudb~k		1.068801	.0662598	1.07	0.283	.9465135	1.206887
pyperstudl~o		1.041629	.0660507	0.64	0.520	.9198934	1.179474
pyperstudm~e		.9487914	.0283914	-1.76	0.079	.8947456	1.006102
Asian		(omitted)					
Black		.8046679	.6346628	-0.28	0.783	.1714951	3.775563
Latino		1.849137	1.189116	0.96	0.339	.5243132	6.521496
Other		4.37e-06	.0027456	-0.02	0.984	0	.
AsianMale		(omitted)					
BlackMale		6.463895	9.41171	1.28	0.200	.3724844	112.171
LatinoMale		.449914	.4471177	-0.80	0.422	.0641532	3.1553
OtherMale		(omitted)					

-----> region = 4

Iteration 0: log likelihood = -5934.5421  
Iteration 1: log likelihood = -5548.3683  
Iteration 2: log likelihood = -5524.0425  
Iteration 3: log likelihood = -5523.8868  
Iteration 4: log likelihood = -5523.8868

Multinomial logistic regression	Number of obs	=	5839
	LR chi2(38)	=	821.31
	Prob > chi2	=	0.0000
Log likelihood = -5523.8868	Pseudo R2	=	0.0692

-----							
	admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----							
1			(base outcome)				
-----							
2							
tenure		1.013009	.0054077	2.42	0.015	1.002465	1.023664
exper		1.04246	.0069114	6.27	0.000	1.029001	1.056094
male		1.14115	.099977	1.51	0.132	.9610992	1.35493



age		.9694403	.0049959	-6.02	0.000	.9596979	.9792816
pyaccrati~2		.954291	.0421766	-1.06	0.290	.8751055	1.040642
pyperstudlep		1.00106	.002581	0.41	0.681	.9960143	1.006132
pyperstude~s		.9948904	.0034584	-1.47	0.141	.9881351	1.001692
pyperstudw~e		.9991755	.0046176	-0.18	0.858	.990166	1.008267
pyperstudb~k		1.000927	.0058305	0.16	0.874	.9895643	1.01242
pyperstudl~o		1.003728	.0056955	0.66	0.512	.9926264	1.014953
pyperstudm~e		1.008277	.0053716	1.55	0.122	.9978039	1.01886
Asian		.6243584	.2224329	-1.32	0.186	.3105891	1.25511
Black		.8811342	.0840472	-1.33	0.185	.7308865	1.062268
Latino		1.597776	.1847699	4.05	0.000	1.273742	2.004243
Other		1.036054	.3625686	0.10	0.919	.5217991	2.057129
AsianMale		.5287667	.3860975	-0.87	0.383	.1263949	2.212069
BlackMale		1.634432	.2549565	3.15	0.002	1.203891	2.218945
LatinoMale		.966827	.1831916	-0.18	0.859	.6669075	1.401625
OtherMale		1.977466	1.224216	1.10	0.271	.5876845	6.653862
-----							
3							
tenure		1.026931	.0060808	4.49	0.000	1.015082	1.038919
exper		1.121831	.0102051	12.64	0.000	1.102007	1.142012
male		1.281659	.1409622	2.26	0.024	1.033127	1.589978
age		.948513	.00738	-6.79	0.000	.9341582	.9630883
pyaccrati~2		1.448642	.0826499	6.50	0.000	1.29538	1.620038
pyperstudlep		1.015513	.0032949	4.74	0.000	1.009075	1.021991
pyperstude~s		1.016179	.0044365	3.68	0.000	1.007521	1.024911
pyperstudw~e		1.011025	.0060745	1.82	0.068	.9991892	1.023001
pyperstudb~k		1.005374	.0076962	0.70	0.484	.9904026	1.020572
pyperstudl~o		.9915839	.0072312	-1.16	0.246	.9775118	1.005858
pyperstudm~e		.9992509	.0068458	-0.11	0.913	.9859231	1.012759
Asian		.5596289	.2649636	-1.23	0.220	.2212522	1.415509
Black		.5390793	.069883	-4.77	0.000	.4181269	.6950199
Latino		1.243659	.1811088	1.50	0.134	.9348556	1.654467
Other		.7149802	.3979874	-0.60	0.547	.2401487	2.128667
AsianMale		1.036136	.9487781	0.04	0.969	.1721794	6.235228
BlackMale		1.797626	.3823687	2.76	0.006	1.184793	2.727447
LatinoMale		.9182877	.2316881	-0.34	0.735	.5600372	1.505708
OtherMale		4.123497	3.387593	1.72	0.085	.8240904	20.63273
-----							

-----> region = 5

Iteration 0: log likelihood = -520.49973  
Iteration 1: log likelihood = -471.40607  
Iteration 2: log likelihood = -469.3157  
Iteration 3: log likelihood = -469.23177  
Iteration 4: log likelihood = -469.21648  
Iteration 5: log likelihood = -469.21312  
Iteration 6: log likelihood = -469.21232  
Iteration 7: log likelihood = -469.21214  
Iteration 8: log likelihood = -469.2121  
Iteration 9: log likelihood = -469.2121

Multinomial logistic regression	Number of obs	=	500
	LR chi2(38)	=	102.58
	Prob > chi2	=	0.0000
Log likelihood = -469.2121	Pseudo R2	=	0.0985

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9943295	.0176314	-0.32	0.748	.9603663 1.029494
exper		.9769468	.0232104	-0.98	0.326	.9324982 1.023514
male		1.360801	.3744608	1.12	0.263	.7935305 2.333595

age		1.024692	.0193591	1.29	0.197	.9874431	1.063347
pyaccrati~2		.8410343	.1373349	-1.06	0.289	.6106868	1.158268
pyperstudlep		1.042894	.034156	1.28	0.200	.9780526	1.112034
pyperstude~s		1.005317	.0105443	0.51	0.613	.9848614	1.026197
pyperstudw~e		.9917908	.0546395	-0.15	0.881	.8902785	1.104878
pyperstudb~k		.9820318	.0553461	-0.32	0.748	.8793321	1.096726
pyperstudl~o		.9607022	.0674632	-0.57	0.568	.8371727	1.102459
pyperstudm~e		.9888161	.0288116	-0.39	0.700	.9339287	1.046929
Asian		5.53e+07	1.73e+11	0.01	0.995	0	.
Black		2.001871	.8186083	1.70	0.090	.8981759	4.461807
Latino		.425106	.4729456	-0.77	0.442	.0480292	3.762612
Other		7.388245	8.933319	1.65	0.098	.6907666	79.02258
AsianMale		1.54e-08	.0000994	-0.00	0.998	0	.
BlackMale		2.028515	1.137055	1.26	0.207	.6761603	6.085649
LatinoMale		1.29e-06	.002056	-0.01	0.993	0	.
OtherMale		1991387	6.22e+09	0.00	0.996	0	.
-----							
3							
tenure		1.005907	.017412	0.34	0.734	.9723521	1.040619
exper		1.085206	.029865	2.97	0.003	1.028223	1.145348
male		2.05188	.5830763	2.53	0.011	1.175625	3.581255
age		.9663506	.0236924	-1.40	0.163	.9210124	1.013921
pyaccrati~2		1.898242	.338274	3.60	0.000	1.338641	2.691776
pyperstudlep		.9983752	.0338176	-0.05	0.962	.9342463	1.066906
pyperstude~s		1.009406	.0104455	0.90	0.366	.989139	1.030087
pyperstudw~e		.9873918	.048811	-0.26	0.797	.8962125	1.087848
pyperstudb~k		.9641826	.0490191	-0.72	0.473	.8727385	1.065208
pyperstudl~o		.991625	.0623901	-0.13	0.894	.8765818	1.121767
pyperstudm~e		1.009916	.0318198	0.31	0.754	.9494367	1.074247
Asian		5.431128	29522.67	0.00	1.000	0	.
Black		2.552881	1.214842	1.97	0.049	1.004543	6.487727
Latino		.4553155	.532153	-0.67	0.501	.046075	4.499455
Other		3.922541	6.311546	0.85	0.396	.1674733	91.8733
AsianMale		7412847	4.71e+10	0.00	0.998	0	.
BlackMale		.9984153	.6667318	-0.00	0.998	.2697039	3.696027
LatinoMale		1.469191	2.801043	0.20	0.840	.0350142	61.64698
OtherMale		.0716685	389.5774	-0.00	1.000	0	.
-----							

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1066.3433  
Iteration 1: log likelihood = -998.72215  
Iteration 2: log likelihood = -996.52106  
Iteration 3: log likelihood = -996.37435  
Iteration 4: log likelihood = -996.34989  
Iteration 5: log likelihood = -996.34435  
Iteration 6: log likelihood = -996.34298  
Iteration 7: log likelihood = -996.34272  
Iteration 8: log likelihood = -996.34267  
Iteration 9: log likelihood = -996.34266

Multinomial logistic regression	Number of obs	=	1018
	LR chi2(36)	=	140.00
	Prob > chi2	=	0.0000
Log likelihood = -996.34266	Pseudo R2	=	0.0656

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.017459	.0132491	1.33	0.184	.9918202 1.043762
exper		1.018067	.0162227	1.12	0.261	.9867626 1.050365

male		1.487344	.2640291	2.24	0.025	1.050288	2.106271
age		.9708205	.0124104	-2.32	0.021	.9467988	.9954517
pyacctrati~2		.9334419	.1045743	-0.61	0.539	.749422	1.162648
pyperstudlep		1.026986	.0129478	2.11	0.035	1.00192	1.052679
pyperstude~s		.9837284	.0068471	-2.36	0.018	.9703994	.9972405
pyperstudw~e		1.033339	.0277021	1.22	0.221	.9804452	1.089086
pyperstudb~k		1.035497	.0304192	1.19	0.235	.9775599	1.096867
pyperstudl~o		1.035868	.0283636	1.29	0.198	.9817416	1.092978
pyperstudm~e		.9992942	.0123929	-0.06	0.955	.9752974	1.023881
Asian		.50907	.5980628	-0.57	0.565	.0509066	5.090744
Black		2.393991	.9203624	2.27	0.023	1.126888	5.085859
Latino		.6118377	.266103	-1.13	0.259	.2608724	1.434975
Other		3.250369	4.714772	0.81	0.416	.1893369	55.79946
AsianMale		(omitted)					
BlackMale		1.35478	.9207565	0.45	0.655	.3575708	5.13305
LatinoMale		1.752393	1.188779	0.83	0.408	.4636583	6.623155
OtherMale		2934482	7.23e+09	0.01	0.995	0	.

3							
tenure		1.016961	.0127225	1.34	0.179	.9923282	1.042205
exper		1.087185	.0192079	4.73	0.000	1.050182	1.125491
male		2.066386	.3971857	3.78	0.000	1.41775	3.01178
age		.9625986	.0147314	-2.49	0.013	.9341543	.9919091
pyacctrati~2		1.061821	.1304205	0.49	0.625	.8346425	1.350834
pyperstudlep		1.040167	.0145539	2.81	0.005	1.01203	1.069087
pyperstude~s		1.021616	.0078382	2.79	0.005	1.006369	1.037095
pyperstudw~e		1.029273	.0282947	1.05	0.294	.9752838	1.086251
pyperstudb~k		1.021004	.0300944	0.71	0.481	.9636912	1.081725
pyperstudl~o		.9975068	.0278069	-0.09	0.929	.9444684	1.053524
pyperstudm~e		.9639042	.0181358	-1.95	0.051	.9290062	1.000113
Asian		1.02e-06	.0009699	-0.01	0.988	0	.
Black		1.858605	.8223532	1.40	0.161	.7808489	4.423921
Latino		.3704412	.2389255	-1.54	0.124	.1046442	1.311364
Other		1.66e-06	.0023296	-0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		.5710486	.4746042	-0.67	0.500	.1120033	2.91149
LatinoMale		3.705191	3.171607	1.53	0.126	.6921345	19.83493
OtherMale		105598.5	4.81e+08	0.00	0.998	0	.

-----> region = 7

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -1198.5806  
Iteration 1: log likelihood = -1133.1101  
Iteration 2: log likelihood = -1131.9044  
Iteration 3: log likelihood = -1131.7933  
Iteration 4: log likelihood = -1131.7723  
Iteration 5: log likelihood = -1131.7678  
Iteration 6: log likelihood = -1131.7668  
Iteration 7: log likelihood = -1131.7665  
Iteration 8: log likelihood = -1131.7665  
Iteration 9: log likelihood = -1131.7665

Multinomial logistic regression	Number of obs	=	1167
	LR chi2(34)	=	133.63
	Prob > chi2	=	0.0000
Log likelihood = -1131.7665	Pseudo R2	=	0.0557

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					

tenure		.9896779	.0114619	-0.90	0.370	.9674661	1.0124
exper		1.02581	.0142969	1.83	0.067	.9981674	1.054217
male		2.004591	.3630603	3.84	0.000	1.405598	2.858845
age		.9687467	.0114823	-2.68	0.007	.9465012	.9915151
pyacctrati~2		.9288238	.1097825	-0.62	0.532	.7367589	1.170958
pyperstudlep		.9984581	.0127657	-0.12	0.904	.9737487	1.023795
pyperstude~s		1.004533	.0078504	0.58	0.563	.9892636	1.020038
pyperstudw~e		.8387023	.0530399	-2.78	0.005	.7409304	.9493759
pyperstudb~k		.8370233	.0543702	-2.74	0.006	.7369643	.9506676
pyperstudl~o		.8459239	.0544959	-2.60	0.009	.745582	.9597699
pyperstudm~e		.9725014	.0186499	-1.45	0.146	.9366266	1.00975
Asian		(omitted)					
Black		1.865683	.5553597	2.10	0.036	1.041022	3.343612
Latino		2.260283	1.372712	1.34	0.179	.6874043	7.432134
Other		2.937853	2.482734	1.28	0.202	.5606516	15.39456
AsianMale		(omitted)					
BlackMale		.9955965	.4833365	-0.01	0.993	.3844562	2.578219
LatinoMale		1.83318	1.922604	0.58	0.563	.2346829	14.31954
OtherMale		.3819567	.4832692	-0.76	0.447	.0319915	4.560301
-----							
3							
tenure		.9917954	.0102	-0.80	0.423	.9720039	1.01199
exper		1.039255	.0133713	2.99	0.003	1.013375	1.065795
male		2.064665	.3505049	4.27	0.000	1.48029	2.879736
age		.9845009	.0110293	-1.39	0.163	.9631194	1.006357
pyacctrati~2		1.138976	.1235293	1.20	0.230	.920866	1.408747
pyperstudlep		1.031939	.0136527	2.38	0.017	1.005524	1.059047
pyperstude~s		1.032282	.0072798	4.51	0.000	1.018112	1.046649
pyperstudw~e		.9150604	.0639964	-1.27	0.204	.7978467	1.049494
pyperstudb~k		.8994782	.0641872	-1.48	0.138	.7820752	1.034506
pyperstudl~o		.8789827	.0626599	-1.81	0.070	.7643652	1.010787
pyperstudm~e		.9925162	.0157268	-0.47	0.635	.9621659	1.023824
Asian		(omitted)					
Black		1.420542	.4446945	1.12	0.262	.7691115	2.623729
Latino		.7794758	.6395057	-0.30	0.761	.1561173	3.891834
Other		1.742915	1.623649	0.60	0.551	.2807514	10.82008
AsianMale		(omitted)					
BlackMale		1.292977	.6303378	0.53	0.598	.4973046	3.361701
LatinoMale		6.124211	7.369564	1.51	0.132	.5791052	64.76536
OtherMale		2.90e-07	.0002772	-0.02	0.987	0	.
-----							

-----> region = 8

note: AsianMale omitted because of collinearity  
note: LatinoMale omitted because of collinearity  
Iteration 0: log likelihood = -389.94942  
Iteration 1: log likelihood = -341.89746  
Iteration 2: log likelihood = -340.28353  
Iteration 3: log likelihood = -340.16855  
Iteration 4: log likelihood = -340.14342  
Iteration 5: log likelihood = -340.1377  
Iteration 6: log likelihood = -340.13645  
Iteration 7: log likelihood = -340.13627  
Iteration 8: log likelihood = -340.13624

Multinomial logistic regression	Number of obs	=	372
	LR chi2(34)	=	99.63
	Prob > chi2	=	0.0000
Log likelihood = -340.13624	Pseudo R2	=	0.1277

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

```

2      |
      tenure | 1.054235 .0226363 2.46 0.014 1.01079 1.099549
      exper | .9777401 .0281201 -0.78 0.434 .9241503 1.034438
      male | 3.424603 1.15791 3.64 0.000 1.765249 6.643767
      age | .9928996 .0241115 -0.29 0.769 .946749 1.0413
pyaccrati~2 | 1.429824 .281987 1.81 0.070 .9714277 2.104529
pyperstudlep | 1.035193 .0311882 1.15 0.251 .9758355 1.098162
pyperstude~s | .9609427 .0137964 -2.77 0.006 .9342792 .9883671
pyperstudw~e | 1.077875 .1658317 0.49 0.626 .7972798 1.457224
pyperstudb~k | 1.144184 .177213 0.87 0.384 .8446182 1.549998
pyperstudl~o | 1.112382 .169239 0.70 0.484 .8255649 1.498845
pyperstudm~e | 1.014369 .0289327 0.50 0.617 .9592177 1.072691
      Asian | .5648102 1511.352 -0.00 1.000 0 .
      Black | 2.62731 1.48433 1.71 0.087 .8681904 7.950743
      Latino | 2.779727 4.076239 0.70 0.486 .1569588 49.22873
      Other | 8.35e-06 .0050654 -0.02 0.985 0 .
      AsianMale | (omitted)
      BlackMale | 1.173967 1.518941 0.12 0.901 .0929665 14.8247
      LatinoMale | (omitted)
      OtherMale | 127373.5 3.49e+08 0.00 0.997 0 .
-----

```

```

3      |
      tenure | 1.046738 .0197282 2.42 0.015 1.008777 1.086128
      exper | .9995908 .0264202 -0.02 0.988 .9491265 1.052738
      male | 2.643987 .789517 3.26 0.001 1.472595 4.747176
      age | .9867348 .0226438 -0.58 0.561 .9433372 1.032129
pyaccrati~2 | 1.996686 .3814972 3.62 0.000 1.373013 2.903653
pyperstudlep | 1.023064 .0309526 0.75 0.451 .9641615 1.085564
pyperstude~s | 1.01203 .0123748 0.98 0.328 .9880637 1.036577
pyperstudw~e | 1.249554 .158495 1.76 0.079 .9745127 1.602221
pyperstudb~k | 1.255321 .1610129 1.77 0.076 .9762839 1.614112
pyperstudl~o | 1.225203 .154663 1.61 0.108 .9566586 1.569131
pyperstudm~e | 1.005088 .0291597 0.17 0.861 .9495308 1.063896
      Asian | 576500.5 8.70e+08 0.01 0.993 0 .
      Black | 1.501569 .9271454 0.66 0.510 .4476823 5.036405
      Latino | 6.694762 8.042459 1.58 0.113 .6355964 70.51619
      Other | 13.52832 15.85122 2.22 0.026 1.361086 134.4629
      AsianMale | (omitted)
      BlackMale | 1.620427 2.260495 0.35 0.729 .1052456 24.94911
      LatinoMale | (omitted)
      OtherMale | 340374 5.14e+08 0.01 0.993 0 .
-----

```

-----> region = 9

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -245.09133
Iteration 1: log likelihood = -218.79517
Iteration 2: log likelihood = -217.22142
Iteration 3: log likelihood = -216.86011
Iteration 4: log likelihood = -216.80243
Iteration 5: log likelihood = -216.78794
Iteration 6: log likelihood = -216.78481
Iteration 7: log likelihood = -216.78411
Iteration 8: log likelihood = -216.784
Iteration 9: log likelihood = -216.78397
Iteration 10: log likelihood = -216.78397

```

```

Multinomial logistic regression      Number of obs   =      230
                                      LR chi2(34)        =      56.61
                                      Prob > chi2         =      0.0088
Log likelihood = -216.78397          Pseudo R2        =      0.1155

```

```

-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]

```

1		(base outcome)					
2							
	tenure	.9895404	.0286538	-0.36	0.717	.9349439	1.047325
	exper	.9848909	.0370421	-0.40	0.686	.9149011	1.060235
	male	.5896301	.2530397	-1.23	0.218	.254265	1.367328
	age	.9888908	.0303038	-0.36	0.715	.9312448	1.050105
	pyacctrati~2	1.482966	.4003046	1.46	0.144	.8736996	2.5171
	pyperstudlep	1.101155	.0778375	1.36	0.173	.9586924	1.264787
	pyperstude~s	.9920421	.0191664	-0.41	0.679	.9551789	1.030328
	pyperstudw~e	.7583646	.0846504	-2.48	0.013	.6093474	.9438241
	pyperstudb~k	.7649694	.0907059	-2.26	0.024	.6063358	.9651057
	pyperstudl~o	.75958	.0884876	-2.36	0.018	.6045225	.9544091
	pyperstudm~e	1.01645	.0302252	0.55	0.583	.9589029	1.077451
	Asian	(omitted)					
	Black	3.6023	4.895045	0.94	0.346	.2511368	51.67128
	Latino	1.00e-07	.0003911	-0.00	0.997	0	.
	Other	1.17e+08	6.63e+11	0.00	0.997	0	.
	AsianMale	(omitted)					
	BlackMale	2.839825	5.312017	0.56	0.577	.0726243	111.0456
	LatinoMale	9098331	4.78e+10	0.00	0.998	0	.
	OtherMale	.0796522	473.4925	-0.00	1.000	0	.
3							
	tenure	1.014693	.0218351	0.68	0.498	.9727875	1.058405
	exper	1.045389	.032907	1.41	0.158	.9828418	1.111917
	male	1.479719	.5071591	1.14	0.253	.7558545	2.896811
	age	.9636552	.0275329	-1.30	0.195	.9111749	1.019158
	pyacctrati~2	1.567007	.3527938	2.00	0.046	1.007935	2.43618
	pyperstudlep	1.038636	.0643191	0.61	0.540	.9199224	1.172668
	pyperstude~s	.9895105	.0155558	-0.67	0.502	.9594867	1.020474
	pyperstudw~e	1.055964	.1004295	0.57	0.567	.8763828	1.272343
	pyperstudb~k	1.050786	.1053059	0.49	0.621	.8633963	1.278847
	pyperstudl~o	1.062366	.105142	0.61	0.541	.8750461	1.289785
	pyperstudm~e	1.028421	.0249503	1.16	0.248	.9806636	1.078504
	Asian	(omitted)					
	Black	2.508948	3.819605	0.60	0.546	.1269487	49.58552
	Latino	1.23e-07	.0003654	-0.01	0.996	0	.
	Other	1.614427	13189.74	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	1.11e-07	.000212	-0.01	0.993	0	.
	LatinoMale	5.28e+13	1.87e+17	0.01	0.993	0	.
	OtherMale	2478290	2.07e+10	0.00	0.999	0	.

-----> region = 10

Iteration 0: log likelihood = -4373.9062  
Iteration 1: log likelihood = -4092.4283  
Iteration 2: log likelihood = -4084.1716  
Iteration 3: log likelihood = -4084.1285  
Iteration 4: log likelihood = -4084.1285

Multinomial logistic regression      Number of obs      =      4175  
LR chi2(38)      =      579.56  
Prob > chi2      =      0.0000  
Log likelihood = -4084.1285      Pseudo R2      =      0.0663

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				
2					
tenure	1.013039	.006489	2.02	0.043	1.000401 1.025838

exper		1.042585	.0082308	5.28	0.000	1.026578	1.058843
male		1.226898	.1241013	2.02	0.043	1.006258	1.495919
age		.9636943	.0060126	-5.93	0.000	.9519816	.9755511
pyacctrati~2		.9364388	.0493346	-1.25	0.213	.8445696	1.038301
pyperstudlep		1.000383	.0039463	0.10	0.923	.9926778	1.008147
pyperstude~s		.9854015	.0038844	-3.73	0.000	.9778176	.9930442
pyperstudw~e		.9928375	.0065247	-1.09	0.274	.9801312	1.005708
pyperstudb~k		.9965551	.007462	-0.46	0.645	.9820367	1.011288
pyperstudl~o		1.000849	.0074464	0.11	0.909	.98636	1.015551
pyperstudm~e		1.002402	.0072822	0.33	0.741	.9882307	1.016777
Asian		1.094932	.4993786	0.20	0.842	.4478882	2.676731
Black		1.340122	.1578432	2.49	0.013	1.063868	1.688112
Latino		1.537418	.2406786	2.75	0.006	1.131198	2.089514
Other		.7600669	.306055	-0.68	0.496	.345223	1.673416
AsianMale		1.111588	.7960901	0.15	0.883	.2731089	4.524306
BlackMale		1.540072	.2996872	2.22	0.026	1.051727	2.255168
LatinoMale		1.431128	.3692873	1.39	0.165	.8630477	2.373132
OtherMale		1.025151	.6626745	0.04	0.969	.2887731	3.639306

3							
tenure		1.01837	.0064853	2.86	0.004	1.005738	1.031161
exper		1.085475	.0096009	9.27	0.000	1.06682	1.104456
male		1.323414	.1472805	2.52	0.012	1.064062	1.64598
age		.9713082	.0072024	-3.93	0.000	.9572938	.9855277
pyacctrati~2		1.402376	.0851915	5.57	0.000	1.244961	1.579695
pyperstudlep		1.018911	.0046399	4.11	0.000	1.009858	1.028046
pyperstude~s		1.016823	.004196	4.04	0.000	1.008632	1.02508
pyperstudw~e		1.000697	.0072126	0.10	0.923	.9866598	1.014933
pyperstudb~k		.990489	.0082978	-1.14	0.254	.9743584	1.006887
pyperstudl~o		.9735365	.007847	-3.33	0.001	.9582776	.9890385
pyperstudm~e		.9851646	.0086178	-1.71	0.088	.968418	1.002201
Asian		1.013754	.5931561	0.02	0.981	.3220236	3.191375
Black		.8967542	.12948	-0.75	0.450	.6757261	1.19008
Latino		1.285115	.224532	1.44	0.151	.912478	1.80993
Other		.4530365	.2394136	-1.50	0.134	.1608065	1.276329
AsianMale		.3745436	.4591537	-0.80	0.423	.0338854	4.139921
BlackMale		1.667631	.4001674	2.13	0.033	1.041943	2.669043
LatinoMale		1.52143	.4470162	1.43	0.153	.8553769	2.706114
OtherMale		2.090181	1.605673	0.96	0.337	.4637594	9.420524

-----> region = 11

Iteration 0: log likelihood = -3234.1188  
Iteration 1: log likelihood = -3038.4671  
Iteration 2: log likelihood = -3033.015  
Iteration 3: log likelihood = -3032.9928  
Iteration 4: log likelihood = -3032.9928

Multinomial logistic regression	Number of obs	=	3041
	LR chi2(38)	=	402.25
	Prob > chi2	=	0.0000
Log likelihood = -3032.9928	Pseudo R2	=	0.0622

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.020694	.0077456	2.70	0.007	1.005625 1.035988
exper		1.021342	.0089264	2.42	0.016	1.003995 1.038988
male		1.245049	.129454	2.11	0.035	1.015506 1.526476
age		.9718891	.0066946	-4.14	0.000	.9588561 .9850992
pyacctrati~2		1.086523	.0688507	1.31	0.190	.9596214 1.230206
pyperstudlep		.9915569	.0050327	-1.67	0.095	.981742 1.00147

pyperstude~s		1.002513	.0042115	0.60	0.550	.9942923	1.010801
pyperstudw~e		1.004029	.0094742	0.43	0.670	.9856303	1.022771
pyperstudb~k		1.004549	.0110447	0.41	0.680	.9831332	1.026431
pyperstudl~o		1.013526	.0105357	1.29	0.196	.9930855	1.034387
pyperstudm~e		.9916229	.006132	-1.36	0.174	.979677	1.003715
Asian		3.323203	2.380854	1.68	0.094	.8160702	13.53276
Black		1.510989	.2462832	2.53	0.011	1.097791	2.079709
Latino		1.581208	.3287542	2.20	0.028	1.051993	2.37665
Other		.5667043	.238538	-1.35	0.177	.2483519	1.29314
AsianMale		.2554911	.2476306	-1.41	0.159	.0382263	1.707611
BlackMale		1.743006	.4956414	1.95	0.051	.9982769	3.043313
LatinoMale		.9477156	.3081999	-0.17	0.869	.5010312	1.792632
OtherMale		1.106085	.8198723	0.14	0.892	.258729	4.728588

3							
tenure		1.020932	.0077491	2.73	0.006	1.005856	1.036233
exper		1.091056	.0110507	8.60	0.000	1.06961	1.112931
male		1.626469	.1905097	4.15	0.000	1.292837	2.046198
age		.9676572	.0083989	-3.79	0.000	.9513349	.9842595
pyacctrati~2		1.535366	.1130035	5.83	0.000	1.329117	1.77362
pyperstudlep		1.007454	.0061305	1.22	0.222	.9955095	1.019541
pyperstude~s		1.029637	.0048677	6.18	0.000	1.020141	1.039222
pyperstudw~e		1.025139	.0116735	2.18	0.029	1.002513	1.048276
pyperstudb~k		1.015598	.0135956	1.16	0.248	.9892976	1.042597
pyperstudl~o		1.005865	.0125825	0.47	0.640	.9815041	1.030832
pyperstudm~e		.9767999	.0079972	-2.87	0.004	.9612509	.9926005
Asian		1.403216	1.349969	0.35	0.725	.2129233	9.247534
Black		1.191257	.2522847	0.83	0.409	.7865708	1.804152
Latino		1.106512	.284499	0.39	0.694	.6684992	1.831519
Other		.6566068	.3170916	-0.87	0.384	.2548234	1.691888
AsianMale		.4116241	.53341	-0.68	0.493	.0324681	5.218488
BlackMale		1.496256	.5570912	1.08	0.279	.7212467	3.104047
LatinoMale		1.393657	.5396709	0.86	0.391	.6524424	2.976938
OtherMale		1.39326	1.092346	0.42	0.672	.2996896	6.477278

-----> region = 12

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -1011.5962
Iteration 1: log likelihood = -941.26238
Iteration 2: log likelihood = -938.83184
Iteration 3: log likelihood = -938.57522
Iteration 4: log likelihood = -938.51396
Iteration 5: log likelihood = -938.50141
Iteration 6: log likelihood = -938.49877
Iteration 7: log likelihood = -938.49817
Iteration 8: log likelihood = -938.49802
Iteration 9: log likelihood = -938.49799
```

Multinomial logistic regression	Number of obs	=	948
	LR chi2(36)	=	146.20
	Prob > chi2	=	0.0000
Log likelihood = -938.49799	Pseudo R2	=	0.0723

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.014691	.0125229	1.18	0.237	.990441 1.039535
exper		1.003434	.0158312	0.22	0.828	.9728801 1.034947
male		.9424286	.1826342	-0.31	0.760	.6446043 1.377856
age		.9879421	.0122571	-0.98	0.328	.9642084 1.01226
pyacctrati~2		1.010018	.1135477	0.09	0.929	.8102813 1.258991



pyperstudlep		1.008723	.0153096	0.57	0.567	.9791583	1.03918
pyperstude~s		.9969179	.0073702	-0.42	0.676	.9825768	1.011468
pyperstudw~e		.8745876	.0509367	-2.30	0.021	.7802407	.9803428
pyperstudb~k		.8856487	.0542829	-1.98	0.048	.7853981	.9986955
pyperstudl~o		.8909788	.0524013	-1.96	0.050	.7939725	.9998372
pyperstudm~e		1.004442	.0111457	0.40	0.690	.9828329	1.026526
Asian		1.13e+07	1.91e+10	0.01	0.992	0	.
Black		.9070936	.27044	-0.33	0.744	.5056799	1.627153
Latino		.3756222	.1804606	-2.04	0.042	.1464899	.9631524
Other		1.26e-06	.0007263	-0.02	0.981	0	.
AsianMale		(omitted)					
BlackMale		2.507752	1.227694	1.88	0.060	.9606626	6.546336
LatinoMale		3.865709	2.647192	1.97	0.048	1.010035	14.79524
OtherMale		2.3748	2298.826	0.00	0.999	0	.
-----							
3							
tenure		.9964255	.0114218	-0.31	0.755	.9742887	1.019065
exper		1.078372	.0187269	4.34	0.000	1.042285	1.115707
male		1.267592	.247706	1.21	0.225	.8642562	1.85916
age		.96361	.0143233	-2.49	0.013	.9359417	.9920961
pyacctrati~2		1.492119	.172319	3.47	0.001	1.189875	1.871136
pyperstudlep		1.034198	.0160966	2.16	0.031	1.003126	1.066233
pyperstude~s		1.007699	.0081033	0.95	0.340	.9919417	1.023708
pyperstudw~e		1.082597	.0697272	1.23	0.218	.9542086	1.228261
pyperstudb~k		1.092412	.0738584	1.31	0.191	.9568331	1.247201
pyperstudl~o		1.071533	.0703214	1.05	0.292	.9422012	1.218617
pyperstudm~e		.9867483	.0140354	-0.94	0.348	.9596192	1.014644
Asian		1.645218	4599.691	0.00	1.000	0	.
Black		.7294865	.2648219	-0.87	0.385	.3581045	1.48602
Latino		.2067048	.1405347	-2.32	0.020	.0545299	.7835495
Other		1.552981	1.264935	0.54	0.589	.3146624	7.664561
AsianMale		(omitted)					
BlackMale		1.650344	.9540627	0.87	0.386	.5314901	5.124524
LatinoMale		4.089537	3.653976	1.58	0.115	.7097916	23.56229
OtherMale		1.227603	1.546593	0.16	0.871	.1039138	14.50249
-----							

-----> region = 13

```
Iteration 0: log likelihood = -1898.2923
Iteration 1: log likelihood = -1811.9548
Iteration 2: log likelihood = -1810.5043
Iteration 3: log likelihood = -1810.4437
Iteration 4: log likelihood = -1810.4338
Iteration 5: log likelihood = -1810.4315
Iteration 6: log likelihood = -1810.431
Iteration 7: log likelihood = -1810.4309
Iteration 8: log likelihood = -1810.4308
```

Multinomial logistic regression	Number of obs	=	1755
	LR chi2(38)	=	175.72
	Prob > chi2	=	0.0000
Log likelihood = -1810.4308	Pseudo R2	=	0.0463

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.001438	.0097137	0.15	0.882	.9825795 1.020659
exper		.984541	.0100897	-1.52	0.128	.9649628 1.004516
male		1.036305	.1479433	0.25	0.803	.7833752 1.370899
age		.9866861	.008279	-1.60	0.110	.9705923 1.003047
pyacctrati~2		1.047691	.0807148	0.60	0.545	.900857 1.218457
pyperstudlep		1.003271	.0061098	0.54	0.592	.9913676 1.015318

pyperstude~s		.9924534	.0059599	-1.26	0.207	.9808406	1.004204
pyperstudw~e		.9814751	.0136984	-1.34	0.180	.9549906	1.008694
pyperstudb~k		.9816688	.0171477	-1.06	0.290	.9486287	1.01586
pyperstudl~o		.9895004	.0147458	-0.71	0.479	.9610171	1.018828
pyperstudm~e		1.006325	.0085007	0.75	0.455	.9898008	1.023124
Asian		2.778481	3.425448	0.83	0.407	.2479732	31.13224
Black		2.0268	.4886006	2.93	0.003	1.263606	3.250948
Latino		1.342594	.266045	1.49	0.137	.9104855	1.979779
Other		.5713726	.4084567	-0.78	0.434	.1407413	2.319622
AsianMale		2.00e-06	.0007972	-0.03	0.974	0	.
BlackMale		.939414	.404778	-0.15	0.885	.4037275	2.185877
LatinoMale		1.229977	.3857322	0.66	0.509	.6651998	2.274268
OtherMale		.7443288	.8157448	-0.27	0.788	.0868747	6.377295

3							
tenure		1.019476	.0095869	2.05	0.040	1.000858	1.03844
exper		1.036606	.0120553	3.09	0.002	1.013245	1.060505
male		.9552924	.1538136	-0.28	0.776	.6967591	1.309755
age		.9896361	.0100938	-1.02	0.307	.970049	1.009619
pyacctrati~2		1.51638	.1304993	4.84	0.000	1.281014	1.794991
pyperstudlep		1.000446	.0066493	0.07	0.946	.9874986	1.013564
pyperstude~s		1.010632	.0064648	1.65	0.098	.9980403	1.023382
pyperstudw~e		.9880739	.0152253	-0.78	0.436	.958679	1.01837
pyperstudb~k		.9744982	.0189966	-1.33	0.185	.9379678	1.012451
pyperstudl~o		.9851143	.0160766	-0.92	0.358	.9541033	1.017133
pyperstudm~e		.9959463	.0104666	-0.39	0.699	.975642	1.016673
Asian		7.432844	8.239043	1.81	0.070	.8465027	65.2652
Black		1.542986	.4483106	1.49	0.135	.8730673	2.726945
Latino		1.035168	.2317692	0.15	0.877	.6674682	1.605428
Other		1.125445	.7501196	0.18	0.859	.3047826	4.155833
AsianMale		.2422812	.4494216	-0.76	0.445	.0063882	9.188892
BlackMale		.8530942	.4786877	-0.28	0.777	.2840346	2.562257
LatinoMale		1.737271	.6232467	1.54	0.124	.8600001	3.509429
OtherMale		.3280092	.4208349	-0.87	0.385	.0265336	4.054853

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -267.90082  
Iteration 1: log likelihood = -225.90264  
Iteration 2: log likelihood = -221.34489  
Iteration 3: log likelihood = -220.86067  
Iteration 4: log likelihood = -220.82393  
Iteration 5: log likelihood = -220.8173  
Iteration 6: log likelihood = -220.81582  
Iteration 7: log likelihood = -220.81546  
Iteration 8: log likelihood = -220.81538  
Iteration 9: log likelihood = -220.81537

Multinomial logistic regression	Number of obs	=	274
	LR chi2(30)	=	94.17
	Prob > chi2	=	0.0000
Log likelihood = -220.81537	Pseudo R2	=	0.1758

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9798442	.0289864	-0.69	0.491	.9246476 1.038336
exper		1.003243	.039827	0.08	0.935	.9281427 1.084419

male		.7774041	.3512143	-0.56	0.577	.3206952	1.884522
age		1.058964	.0392257	1.55	0.122	.9848071	1.138704
pyacctrati~2		.7270323	.2210415	-1.05	0.294	.4006459	1.319309
pyperstudlep		1.049853	.1134823	0.45	0.653	.8494137	1.297592
pyperstude~s		.9595736	.0218207	-1.81	0.070	.9177449	1.003309
pyperstudw~e		.8698693	.1777645	-0.68	0.495	.5827797	1.298385
pyperstudb~k		.9050377	.2183624	-0.41	0.679	.5640187	1.452245
pyperstudl~o		.9015275	.1821691	-0.51	0.608	.6067071	1.339612
pyperstudm~e		1.021439	.0444044	0.49	0.626	.938013	1.112286
Asian		(omitted)					
Black		7032195	8.12e+09	0.01	0.989	0	.
Latino		.0000115	.0089282	-0.01	0.988	0	.
Other		.0000157	.0123736	-0.01	0.989	0	.
AsianMale		(omitted)					
BlackMale		(omitted)					
LatinoMale		1489346	1.15e+09	0.02	0.985	0	.
OtherMale		(omitted)					
-----							
3							
tenure		1.004961	.0214022	0.23	0.816	.963877	1.047796
exper		1.026324	.0302969	0.88	0.379	.9686288	1.087457
male		3.279299	1.101919	3.53	0.000	1.697298	6.335839
age		1.034934	.0281191	1.26	0.206	.9812632	1.09154
pyacctrati~2		1.337493	.2751107	1.41	0.157	.8937305	2.001595
pyperstudlep		1.127521	.0678178	2.00	0.046	1.002137	1.268594
pyperstude~s		1.050603	.0149795	3.46	0.001	1.02165	1.080377
pyperstudw~e		1.010601	.1656669	0.06	0.949	.7328967	1.39353
pyperstudb~k		.9661114	.1757885	-0.19	0.850	.676312	1.38009
pyperstudl~o		.9834907	.1602341	-0.10	0.919	.7146432	1.353478
pyperstudm~e		.9746023	.0271952	-0.92	0.357	.9227319	1.029388
Asian		(omitted)					
Black		2258737	2.61e+09	0.01	0.990	0	.
Latino		4.415253	5.974982	1.10	0.272	.3112141	62.64003
Other		1.607542	2.081275	0.37	0.714	.1270916	20.33329
AsianMale		(omitted)					
BlackMale		(omitted)					
LatinoMale		.9223006	1.672559	-0.04	0.964	.0263785	32.24747
OtherMale		(omitted)					
-----							

-----> region = 15

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -284.16402  
Iteration 1: log likelihood = -251.33827  
Iteration 2: log likelihood = -246.04347  
Iteration 3: log likelihood = -245.85281  
Iteration 4: log likelihood = -245.83466  
Iteration 5: log likelihood = -245.83084  
Iteration 6: log likelihood = -245.82997  
Iteration 7: log likelihood = -245.82984  
Iteration 8: log likelihood = -245.82982

Multinomial logistic regression	Number of obs	=	274
	LR chi2(30)	=	76.67
	Prob > chi2	=	0.0000
Log likelihood = -245.82982	Pseudo R2	=	0.1349

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1	(base outcome)				

```

2      |
      tenure | .9978737 .0359058 -0.06 0.953 .9299237 1.070789
      exper | .9428781 .0348994 -1.59 0.112 .8768987 1.013822
      male | .6335802 .3185819 -0.91 0.364 .2364796 1.697499
      age | .9949494 .0288265 -0.17 0.861 .9400247 1.053083
pyacctrati~2 | .4902826 .1451977 -2.41 0.016 .2743862 .8760535
pyperstudlep | .9634207 .0322308 -1.11 0.265 .9022761 1.028709
pyperstude~s | 1.025416 .0181482 1.42 0.156 .9904565 1.06161
pyperstudw~e | .6310843 .1308879 -2.22 0.026 .4202878 .9476065
pyperstudb~k | .7483239 .1729364 -1.25 0.210 .4757496 1.177066
pyperstudl~o | .6367529 .1318503 -2.18 0.029 .4243415 .9554904
pyperstudm~e | .9979876 .0289032 -0.07 0.945 .9429162 1.056276
      Asian | (omitted)
      Black | 6.561048 8.728803 1.41 0.157 .4836562 89.00404
      Latino | 3.699127 2.129762 2.27 0.023 1.196802 11.43342
      Other | 3.59e-11 2.15e-08 -0.04 0.968 0 .
      AsianMale | (omitted)
      BlackMale | (omitted)
      LatinoMale | 2.314767 2.03653 0.95 0.340 .4126872 12.98355
      OtherMale | (omitted)

```

```

3      |
      tenure | .9675466 .0204375 -1.56 0.118 .9283077 1.008444
      exper | 1.044216 .0278616 1.62 0.105 .9910119 1.100277
      male | .8086954 .2732792 -0.63 0.530 .4170055 1.568296
      age | .9625078 .0235886 -1.56 0.119 .9173678 1.009869
pyacctrati~2 | 1.501275 .3056559 2.00 0.046 1.007297 2.237499
pyperstudlep | 1.006978 .0228894 0.31 0.760 .9631003 1.052855
pyperstude~s | 1.017996 .0118203 1.54 0.125 .9950902 1.041429
pyperstudw~e | 1.101798 .1825709 0.59 0.559 .7962615 1.524574
pyperstudb~k | 1.088695 .2181111 0.42 0.671 .7351483 1.612269
pyperstudl~o | 1.09554 .1809033 0.55 0.581 .7926324 1.514205
pyperstudm~e | .991471 .0179645 -0.47 0.636 .956879 1.027314
      Asian | (omitted)
      Black | 2.098781 3.054856 0.51 0.611 .1210627 36.38512
      Latino | .6935965 .349677 -0.73 0.468 .2582101 1.863119
      Other | 3.671928 7.764376 0.62 0.538 .0582115 231.6218
      AsianMale | (omitted)
      BlackMale | (omitted)
      LatinoMale | 1.529708 1.276587 0.51 0.610 .2980272 7.85166
      OtherMale | (omitted)

```

-----> region = 16

```

note: AsianMale omitted because of collinearity
note: OtherMale omitted because of collinearity
Iteration 0: log likelihood = -549.61534
Iteration 1: log likelihood = -496.79171
Iteration 2: log likelihood = -495.40287
Iteration 3: log likelihood = -495.26537
Iteration 4: log likelihood = -495.2332
Iteration 5: log likelihood = -495.22659
Iteration 6: log likelihood = -495.2252
Iteration 7: log likelihood = -495.22489
Iteration 8: log likelihood = -495.22481
Iteration 9: log likelihood = -495.22479

```

```

Multinomial logistic regression      Number of obs   =      508
                                     LR chi2(34)        =     108.78
                                     Prob > chi2         =      0.0000
Log likelihood = -495.22479          Pseudo R2        =      0.0990

```

```

-----+-----
      admin |          RRR      Std. Err.      z    P>|z|      [95% Conf. Interval]
-----+-----

```

1		(base outcome)					
2	tenure	1.075734	.021029	3.73	0.000	1.035298	1.11775
	exper	.9377455	.0247231	-2.44	0.015	.8905198	.9874757
	male	.9137029	.238119	-0.35	0.729	.5482446	1.522775
	age	.9876258	.0205818	-0.60	0.550	.9480989	1.028801
	pyacctrati~2	1.203211	.1768174	1.26	0.208	.9020978	1.604832
	pyperstudlep	1.033954	.0166072	2.08	0.038	1.001912	1.067022
	pyperstude~s	.9905635	.009836	-0.95	0.340	.9714716	1.010031
	pyperstudw~e	1.010004	.0312923	0.32	0.748	.9504976	1.073237
	pyperstudb~k	1.043234	.0409682	1.08	0.281	.9659501	1.126701
	pyperstudl~o	1.012537	.0298521	0.42	0.673	.9556866	1.07277
	pyperstudm~e	1.001799	.0129163	0.14	0.889	.9768011	1.027437
	Asian	1.09e-06	.0022903	-0.01	0.995	0	.
	Black	1.325133	1.121541	0.33	0.739	.2522525	6.961189
	Latino	2.225765	1.472878	1.21	0.227	.608422	8.142421
	Other	1198018	1.10e+09	0.02	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	315726.3	3.06e+08	0.01	0.990	0	.
	LatinoMale	.8527827	.7867569	-0.17	0.863	.1398107	5.201594
	OtherMale	(omitted)					
3	tenure	1.029934	.0163732	1.86	0.064	.9983379	1.06253
	exper	1.020756	.0227232	0.92	0.356	.9771775	1.066279
	male	1.831159	.4590686	2.41	0.016	1.120292	2.993098
	age	.9821655	.0201513	-0.88	0.380	.9434533	1.022466
	pyacctrati~2	1.736445	.2490942	3.85	0.000	1.310858	2.300204
	pyperstudlep	1.051279	.0163892	3.21	0.001	1.019643	1.083897
	pyperstude~s	1.015987	.0085928	1.88	0.061	.999284	1.032969
	pyperstudw~e	1.089487	.0417418	2.24	0.025	1.010671	1.17445
	pyperstudb~k	1.091615	.0502833	1.90	0.057	.9973792	1.194754
	pyperstudl~o	1.062223	.0393648	1.63	0.103	.9878045	1.142247
	pyperstudm~e	.9880854	.0128954	-0.92	0.358	.9631315	1.013686
	Asian	4.65e-07	.0009001	-0.01	0.994	0	.
	Black	1.448465	1.461779	0.37	0.714	.2003929	10.46969
	Latino	4.648457	3.074572	2.32	0.020	1.271478	16.99451
	Other	1090328	1.00e+09	0.02	0.988	0	.
	AsianMale	(omitted)					
	BlackMale	150938.3	1.46e+08	0.01	0.990	0	.
	LatinoMale	.348944	.3314966	-1.11	0.268	.0542144	2.245933
	OtherMale	(omitted)					

-----> region = 17

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -585.39273  
Iteration 1: log likelihood = -514.36339  
Iteration 2: log likelihood = -511.32224  
Iteration 3: log likelihood = -511.07193  
Iteration 4: log likelihood = -511.02687  
Iteration 5: log likelihood = -511.01675  
Iteration 6: log likelihood = -511.01447  
Iteration 7: log likelihood = -511.01392  
Iteration 8: log likelihood = -511.01381  
Iteration 9: log likelihood = -511.01379

Multinomial logistic regression	Number of obs	=	574
	LR chi2(34)	=	148.76
	Prob > chi2	=	0.0000
Log likelihood = -511.01379	Pseudo R2	=	0.1271

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
tenure		1.017576	.0204931	0.87	0.387	.9781925	1.058545
exper		1.038747	.0287092	1.38	0.169	.9839749	1.096568
male		1.968013	.5448523	2.45	0.014	1.143851	3.385996
age		.961897	.0219451	-1.70	0.089	.9198329	1.005885
pyacctrati~2		.9011396	.148426	-0.63	0.527	.6525163	1.244494
pyperstudlep		1.029457	.0279402	1.07	0.285	.9761261	1.085701
pyperstude~s		.9983891	.0131476	-0.12	0.903	.97295	1.024493
pyperstudw~e		1.01328	.0720383	0.19	0.853	.8814833	1.164783
pyperstudb~k		.9998165	.0740363	-0.00	0.998	.8647467	1.155984
pyperstudl~o		1.000197	.0722857	0.00	0.998	.8680966	1.1524
pyperstudm~e		1.046632	.0170641	2.80	0.005	1.013715	1.080617
Asian		(omitted)					
Black		1.351923	1.181	0.35	0.730	.2439872	7.490948
Latino		.9745376	.5361687	-0.05	0.963	.3315023	2.864907
Other		3.41e-07	.0007281	-0.01	0.994	0	.
AsianMale		(omitted)					
BlackMale		1949761	1.65e+09	0.02	0.986	0	.
LatinoMale		2.695024	1.959296	1.36	0.173	.6482385	11.20444
OtherMale		3.889742	10743.01	0.00	1.000	0	.
3							
tenure		1.028295	.017234	1.66	0.096	.9950661	1.062634
exper		1.050239	.0234119	2.20	0.028	1.00534	1.097142
male		3.517637	.8851644	5.00	0.000	2.148117	5.760286
age		.9962208	.018805	-0.20	0.841	.9600371	1.033768
pyacctrati~2		1.117658	.1658843	0.75	0.454	.8355501	1.495014
pyperstudlep		1.023195	.0225488	1.04	0.298	.979941	1.068358
pyperstude~s		1.051582	.0128245	4.12	0.000	1.026745	1.077021
pyperstudw~e		1.087117	.0810503	1.12	0.263	.9393222	1.258165
pyperstudb~k		1.04719	.0802132	0.60	0.547	.9012075	1.21682
pyperstudl~o		1.061785	.0797176	0.80	0.425	.9164937	1.23011
pyperstudm~e		.9481244	.0195497	-2.58	0.010	.9105716	.9872259
Asian		(omitted)					
Black		.5579252	.6632587	-0.49	0.624	.054284	5.734298
Latino		.9187012	.4657149	-0.17	0.867	.3401549	2.481257
Other		5.74e-07	.0008841	-0.01	0.993	0	.
AsianMale		(omitted)					
BlackMale		6718908	5.67e+09	0.02	0.985	0	.
LatinoMale		1.492201	1.071903	0.56	0.577	.3650717	6.099255
OtherMale		1896236	2.92e+09	0.01	0.993	0	.

-----> region = 18

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -411.3358  
Iteration 1: log likelihood = -373.02568  
Iteration 2: log likelihood = -371.70983  
Iteration 3: log likelihood = -371.63633  
Iteration 4: log likelihood = -371.61963  
Iteration 5: log likelihood = -371.61614  
Iteration 6: log likelihood = -371.61542  
Iteration 7: log likelihood = -371.61526  
Iteration 8: log likelihood = -371.61522  
Iteration 9: log likelihood = -371.61521

Multinomial logistic regression

Number of obs = 381  
LR chi2(36) = 79.44  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0966

Log likelihood = -371.61521

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013648	.0231411	0.59	0.553	.9692918	1.060033
	exper	.9632035	.0266895	-1.35	0.176	.9122881	1.016961
	male	.9859132	.3562974	-0.04	0.969	.4855394	2.001948
	age	1.002747	.020836	0.13	0.895	.9627294	1.044428
pyacctrati~2		.5949473	.1194503	-2.59	0.010	.4014013	.8818163
pyperstudlep		.9696735	.0179591	-1.66	0.096	.9351055	1.005519
pyperstude~s		1.02011	.0149752	1.36	0.175	.9911777	1.049888
pyperstudw~e		.8003533	.0982238	-1.81	0.070	.6292417	1.017996
pyperstudb~k		.8081143	.1015977	-1.69	0.090	.6316232	1.033921
pyperstudl~o		.8036019	.0984584	-1.78	0.074	.6320488	1.021719
pyperstudm~e		.9819771	.0219396	-0.81	0.416	.9399043	1.025933
	Asian	3.70e-07	.0007616	-0.01	0.994	0	.
	Black	10.33616	11.73645	2.06	0.040	1.116456	95.69235
	Latino	2.438403	1.000362	2.17	0.030	1.091183	5.448957
	Other	1.038348	3236.273	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.2142318	.3040844	-1.09	0.278	.013264	3.460146
	LatinoMale	.8285467	.5255862	-0.30	0.767	.238979	2.872594
	OtherMale	1.340703	5909.486	0.00	1.000	0	.
3							
	tenure	1.031422	.0199867	1.60	0.110	.9929833	1.071348
	exper	1.008583	.0265051	0.33	0.745	.9579488	1.061893
	male	1.551978	.496473	1.37	0.169	.8290652	2.905242
	age	.9790762	.0202899	-1.02	0.308	.9401054	1.019662
pyacctrati~2		1.153877	.201352	0.82	0.412	.8196421	1.624405
pyperstudlep		.9895798	.0148353	-0.70	0.485	.9609261	1.019088
pyperstude~s		1.049672	.014498	3.51	0.000	1.021637	1.078476
pyperstudw~e		.9226823	.1051931	-0.71	0.480	.7379186	1.153708
pyperstudb~k		.889072	.105726	-0.99	0.323	.7042297	1.122431
pyperstudl~o		.890009	.1021025	-1.02	0.310	.7107942	1.11441
pyperstudm~e		.9643015	.0235166	-1.49	0.136	.9192939	1.011513
	Asian	2.62e-07	.0004649	-0.01	0.993	0	.
	Black	4.659393	5.896573	1.22	0.224	.3900438	55.66027
	Latino	.9510901	.3756647	-0.13	0.899	.438542	2.062681
	Other	2758880	5.31e+09	0.01	0.994	0	.
	AsianMale	(omitted)					
	BlackMale	.0834915	.1459939	-1.42	0.156	.0027116	2.570753
	LatinoMale	2.004166	1.190115	1.17	0.242	.6258533	6.417927
	OtherMale	2.318128	6306.523	0.00	1.000	0	.

-----> region = 19

note: AsianMale omitted because of collinearity

```
Iteration 0: log likelihood = -928.5307
Iteration 1: log likelihood = -865.78144
Iteration 2: log likelihood = -863.36046
Iteration 3: log likelihood = -863.2166
Iteration 4: log likelihood = -863.18764
Iteration 5: log likelihood = -863.18155
Iteration 6: log likelihood = -863.1802
Iteration 7: log likelihood = -863.17986
Iteration 8: log likelihood = -863.1798
Iteration 9: log likelihood = -863.17978
```

Multinomial logistic regression

```
Number of obs   =      863
LR chi2(36)     =     130.70
Prob > chi2     =     0.0000
Pseudo R2      =     0.0704
```

Log likelihood = -863.17978

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.001708	.0137062	0.12	0.901	.9752009	1.028935
	exper	1.052904	.0185367	2.93	0.003	1.017192	1.089869
	male	.9537152	.3039238	-0.15	0.882	.510697	1.781042
	age	.9794101	.0127963	-1.59	0.111	.9546482	1.004814
pyacctrati~2		1.026623	.1190204	0.23	0.821	.8179517	1.288528
pyperstudlep		1.001	.005858	0.17	0.864	.9895837	1.012547
pyperstude~s		.9873686	.0096667	-1.30	0.194	.9686029	1.006498
pyperstudw~e		.9092154	.0869266	-1.00	0.320	.7538529	1.096597
pyperstudb~k		1.006388	.0938577	0.07	0.946	.8382634	1.208231
pyperstudl~o		.9510373	.0808834	-0.59	0.555	.8050168	1.123544
pyperstudm~e		1.004607	.0131816	0.35	0.726	.9791008	1.030778
	Asian	5.09e-07	.0007396	-0.01	0.992	0	.
	Black	1.235789	.5508869	0.47	0.635	.5158194	2.960676
	Latino	1.277152	.3052697	1.02	0.306	.7994384	2.040328
	Other	1.231227	1.172201	0.22	0.827	.1905213	7.956693
	AsianMale	(omitted)					
	BlackMale	1.542923	1.309962	0.51	0.609	.2921883	8.147522
	LatinoMale	1.166904	.4385746	0.41	0.681	.5586187	2.437556
	OtherMale	3.247643	5.109699	0.75	0.454	.1487049	70.92696
3							
	tenure	.983918	.0143232	-1.11	0.265	.9562416	1.012395
	exper	1.147573	.024595	6.42	0.000	1.100366	1.196805
	male	.871084	.3132781	-0.38	0.701	.4304596	1.762737
	age	.9575712	.0165939	-2.50	0.012	.9255939	.9906533
pyacctrati~2		1.340458	.1831483	2.14	0.032	1.02554	1.75208
pyperstudlep		1.016549	.0066329	2.52	0.012	1.003632	1.029633
pyperstude~s		1.018768	.0112294	1.69	0.092	.9969946	1.041017
pyperstudw~e		.9664719	.093099	-0.35	0.723	.8001918	1.167305
pyperstudb~k		.916453	.0934718	-0.86	0.392	.7504014	1.119249
pyperstudl~o		.9233024	.0826389	-0.89	0.373	.7747443	1.100347
pyperstudm~e		1.006656	.0147811	0.45	0.651	.9780988	1.036047
	Asian	5.36e-07	.0009645	-0.01	0.994	0	.
	Black	.3502295	.24416	-1.50	0.132	.089319	1.373289
	Latino	.8749218	.2332631	-0.50	0.616	.5188361	1.475395
	Other	4.52e-06	.0026199	-0.02	0.983	0	.
	AsianMale	(omitted)					
	BlackMale	1.842484	2.510106	0.45	0.654	.1275764	26.60952
	LatinoMale	1.422186	.6154554	0.81	0.416	.6089708	3.321364
	OtherMale	1656231	9.60e+08	0.02	0.980	0	.

-----> region = 20

Iteration 0: log likelihood = -2196.0008  
Iteration 1: log likelihood = -2099.304  
Iteration 2: log likelihood = -2096.7838  
Iteration 3: log likelihood = -2096.646  
Iteration 4: log likelihood = -2096.6156  
Iteration 5: log likelihood = -2096.6092  
Iteration 6: log likelihood = -2096.6081  
Iteration 7: log likelihood = -2096.608  
Iteration 8: log likelihood = -2096.608

Multinomial logistic regression

Number of obs = 2111  
LR chi2(38) = 198.79  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.0453

Log likelihood = -2096.608



	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9909531	.0079038	-1.14	0.255	.9755823	1.006566
	exper	1.024838	.010575	2.38	0.017	1.004319	1.045775
	male	1.377735	.2098659	2.10	0.035	1.022127	1.857062
	age	.9781279	.008379	-2.58	0.010	.9618424	.9946891
pyacctrati~2		.9360916	.0714013	-0.87	0.387	.8061059	1.087038
pyperstudlep		.9928359	.0066626	-1.07	0.284	.9798628	1.005981
pyperstude~s		1.003184	.005376	0.59	0.553	.9927021	1.013776
pyperstudw~e		.9876564	.0289444	-0.42	0.672	.9325249	1.046047
pyperstudb~k		.987306	.0318541	-0.40	0.692	.9268062	1.051755
pyperstudl~o		.9865161	.029782	-0.45	0.653	.9298378	1.046649
pyperstudm~e		1.00332	.0082689	0.40	0.688	.9872434	1.019658
	Asian	.7599913	.6432862	-0.32	0.746	.14465	3.992994
	Black	.883629	.2741398	-0.40	0.690	.4810504	1.623115
	Latino	1.301157	.1875442	1.83	0.068	.9809359	1.725913
	Other	.4977703	.3969028	-0.87	0.382	.1043071	2.375441
	AsianMale	3.744021	5.620907	0.88	0.379	.1974365	70.99848
	BlackMale	2.144035	.986245	1.66	0.097	.8703288	5.281778
	LatinoMale	1.100523	.2480746	0.42	0.671	.7074979	1.711878
	OtherMale	.4905289	.6631668	-0.53	0.598	.0346648	6.941286
3							
	tenure	1.010052	.0083603	1.21	0.227	.993798	1.026571
	exper	1.074197	.0134175	5.73	0.000	1.048218	1.100819
	male	1.251675	.2188806	1.28	0.199	.8884684	1.763362
	age	.9684142	.0105579	-2.94	0.003	.9479406	.9893299
pyacctrati~2		1.266324	.1073757	2.78	0.005	1.072429	1.495275
pyperstudlep		1.012574	.0065673	1.93	0.054	.9997839	1.025528
pyperstude~s		1.01844	.0061453	3.03	0.002	1.006467	1.030556
pyperstudw~e		1.03971	.03527	1.15	0.251	.9728302	1.111188
pyperstudb~k		1.025303	.0382418	0.67	0.503	.9530242	1.103063
pyperstudl~o		1.022505	.0357895	0.64	0.525	.9547109	1.095113
pyperstudm~e		1.005388	.0093958	0.57	0.565	.9871401	1.023973
	Asian	.8382047	.7223251	-0.20	0.838	.1548192	4.538113
	Black	1.582585	.4655354	1.56	0.119	.8891532	2.816811
	Latino	.937137	.1521148	-0.40	0.689	.6817689	1.288158
	Other	.7697246	.6252311	-0.32	0.747	.1566471	3.782235
	AsianMale	5.771252	9.756183	1.04	0.300	.2100542	158.5655
	BlackMale	1.064408	.5439163	0.12	0.903	.3909678	2.897845
	LatinoMale	.9829996	.2662434	-0.06	0.950	.5781056	1.671473
	OtherMale	2.63e-06	.0011489	-0.03	0.977	0	.

by pyschooltype:

```
. bysort pyschooltype: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstudecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2010a==1, rrr iter(20)
```

```
-----> pyschooltype =
no observations
```

```
-----> pyschooltype = B
```

```
note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -396.54563
Iteration 1: log likelihood = -367.65442
```

```

Iteration 2: log likelihood = -359.3985
Iteration 3: log likelihood = -359.00058
Iteration 4: log likelihood = -358.97492
Iteration 5: log likelihood = -358.96955
Iteration 6: log likelihood = -358.96842
Iteration 7: log likelihood = -358.96817
Iteration 8: log likelihood = -358.96811
Iteration 9: log likelihood = -358.9681

```

```

Multinomial logistic regression
Log likelihood = -358.9681
Number of obs   =      448
LR chi2(34)     =      75.16
Prob > chi2     =      0.0001
Pseudo R2      =      0.0948

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013998	.0326581	0.43	0.666	.9519677	1.08007
	exper	1.000207	.0409418	0.01	0.996	.9230966	1.083758
	male	1.20598	.5384082	0.42	0.675	.5027157	2.893063
	age	.9572501	.0315596	-1.33	0.185	.8973506	1.021148
pyacctrati~2		.6320808	.1718304	-1.69	0.092	.3710005	1.076888
pyperstudlep		1.011682	.023226	0.51	0.613	.9671688	1.058244
pyperstude~s		.968894	.0143567	-2.13	0.033	.94116	.9974452
pyperstudw~e		.956388	.0322615	-1.32	0.186	.8952015	1.021757
pyperstudb~k		.9967027	.0340567	-0.10	0.923	.9321388	1.065739
pyperstudl~o		.9867976	.0340428	-0.39	0.700	.9222807	1.055828
pyperstudm~e		.9252961	.0341826	-2.10	0.036	.8606674	.9947779
	Asian	(omitted)					
	Black	1.941736	1.819921	0.71	0.479	.3093064	12.18966
	Latino	2.381812	2.073186	1.00	0.319	.4325148	13.11638
	Other	2.107475	6623.998	0.00	1.000	0	.
	AsianMale	(omitted)					
	BlackMale	.627434	.9807742	-0.30	0.766	.0293097	13.43151
	LatinoMale	.8214405	.8934348	-0.18	0.856	.0974478	6.924371
	OtherMale	1.114718	4169.311	0.00	1.000	0	.
3							
	tenure	.9684036	.0160294	-1.94	0.052	.9374908	1.000336
	exper	1.014031	.0186534	0.76	0.449	.9781226	1.051259
	male	1.380742	.3279532	1.36	0.174	.866833	2.199326
	age	1.000108	.016672	0.01	0.995	.9679596	1.033324
pyacctrati~2		.863851	.1358096	-0.93	0.352	.6347723	1.175601
pyperstudlep		1.003368	.014637	0.23	0.818	.975086	1.03247
pyperstude~s		.9965132	.0089459	-0.39	0.697	.9791329	1.014202
pyperstudw~e		1.012248	.0317884	0.39	0.698	.9518225	1.076509
pyperstudb~k		1.008673	.0325746	0.27	0.789	.9468062	1.074581
pyperstudl~o		1.022763	.0325239	0.71	0.479	.960963	1.088537
pyperstudm~e		.95677	.0206984	-2.04	0.041	.91705	.9982104
	Asian	(omitted)					
	Black	1.684348	1.376794	0.64	0.524	.3393562	8.360029
	Latino	1.615218	1.053936	0.73	0.462	.4495925	5.802875
	Other	2170797	2.44e+09	0.01	0.990	0	.
	AsianMale	(omitted)					
	BlackMale	.5834093	.7052373	-0.45	0.656	.0545787	6.23625
	LatinoMale	.616247	.5273097	-0.57	0.572	.1151859	3.296934
	OtherMale	1.065576	1465.392	0.00	1.000	0	.

```

-----> pyschooltype = E

```

```

Iteration 0: log likelihood = -12760.375
Iteration 1: log likelihood = -12028.949

```

```

Iteration 2:  log likelihood = -12020.827
Iteration 3:  log likelihood = -12020.82
Iteration 4:  log likelihood = -12020.82

```

```

Multinomial logistic regression      Number of obs   =      11771
                                     LR chi2(38)        =      1479.11
                                     Prob > chi2         =       0.0000
Log likelihood = -12020.82           Pseudo R2        =       0.0580

```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)						
2							
	tenure	1.011451	.0042034	2.74	0.006	1.003245	1.019723
	exper	1.024565	.0053006	4.69	0.000	1.014228	1.035007
	male	1.550828	.1300497	5.23	0.000	1.31578	1.827864
	age	.9772505	.0037724	-5.96	0.000	.9698848	.9846723
	pyacctrati~2	.9141863	.033044	-2.48	0.013	.8516622	.9813005
	pyperstudlep	1.004028	.0016616	2.43	0.015	1.000777	1.00729
	pyperstude~s	.9922024	.0020191	-3.85	0.000	.988253	.9961676
	pyperstudw~e	.9976641	.0043842	-0.53	0.595	.9891082	1.006294
	pyperstudb~k	.9937944	.0051113	-1.21	0.226	.9838267	1.003863
	pyperstudl~o	.9997097	.0046586	-0.06	0.950	.9906206	1.008882
	pyperstudm~e	1.009698	.0037458	2.60	0.009	1.002383	1.017067
	Asian	.6272229	.2069898	-1.41	0.158	.328486	1.197642
	Black	1.037619	.0907457	0.42	0.673	.8741689	1.23163
	Latino	1.158701	.0819789	2.08	0.037	1.008668	1.331051
	Other	1.203002	.3010515	0.74	0.460	.7366359	1.964625
	AsianMale	1.301282	.8813031	0.39	0.697	.3450553	4.90743
	BlackMale	1.099401	.2112869	0.49	0.622	.754343	1.602297
	LatinoMale	.7094774	.0976337	-2.49	0.013	.5417535	.9291277
	OtherMale	.4258058	.3154893	-1.15	0.249	.0996636	1.819225
3							
	tenure	1.01571	.0036964	4.28	0.000	1.008491	1.022981
	exper	1.08631	.005408	16.63	0.000	1.075762	1.096961
	male	2.36122	.1819907	11.15	0.000	2.03016	2.746267
	age	.981562	.0039096	-4.67	0.000	.973929	.9892547
	pyacctrati~2	.9374321	.0336713	-1.80	0.072	.873707	1.005805
	pyperstudlep	.9958417	.0016517	-2.51	0.012	.9926098	.9990842
	pyperstude~s	1.003129	.001977	1.59	0.113	.9992614	1.007011
	pyperstudw~e	1.009234	.0047033	1.97	0.049	1.000057	1.018494
	pyperstudb~k	.9967662	.0053665	-0.60	0.547	.9863035	1.00734
	pyperstudl~o	1.004255	.004939	0.86	0.388	.9946215	1.013982
	pyperstudm~e	1.003943	.003787	1.04	0.297	.9965482	1.011393
	Asian	.8324829	.2902429	-0.53	0.599	.4203453	1.648711
	Black	.9747938	.0873428	-0.28	0.776	.8177942	1.161934
	Latino	.9723098	.0706775	-0.39	0.699	.8431999	1.121189
	Other	1.072114	.2862362	0.26	0.794	.6353082	1.809244
	AsianMale	1.114722	.7334801	0.17	0.869	.3069598	4.048105
	BlackMale	.7132908	.140321	-1.72	0.086	.4850823	1.048861
	LatinoMale	.589983	.0803142	-3.88	0.000	.4518203	.7703946
	OtherMale	.5162163	.354407	-0.96	0.335	.1344131	1.98254

```

-----> pyschooltype = M

```

```

Iteration 0:  log likelihood = -6890.908
Iteration 1:  log likelihood = -6563.3591
Iteration 2:  log likelihood = -6554.4387
Iteration 3:  log likelihood = -6554.4049
Iteration 4:  log likelihood = -6554.4049

```

```

Multinomial logistic regression      Number of obs   =      6606

```

Log likelihood = -6554.4049

LR chi2(38) = 673.01  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0488

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.005645	.0048945	1.16	0.247	.9960974	1.015284
exper	1.026391	.0059915	4.46	0.000	1.014715	1.038202
male	1.597812	.1200371	6.24	0.000	1.379045	1.851284
age	.9785576	.0045184	-4.69	0.000	.9697417	.9874537
pyacctrati~2	.8895752	.0404438	-2.57	0.010	.813736	.9724825
pyperstudlep	1.008004	.0035985	2.23	0.026	1.000975	1.015081
pyperstude~s	.995462	.0026787	-1.69	0.091	.9902255	1.000726
pyperstudw~e	.9962285	.0062597	-0.60	0.548	.9840349	1.008573
pyperstudb~k	.9996877	.0072802	-0.04	0.966	.9855201	1.014059
pyperstudl~o	.997244	.0067889	-0.41	0.685	.9840264	1.010639
pyperstudm~e	1.005927	.0045562	1.30	0.192	.9970367	1.014897
Asian	1.393171	.6219881	0.74	0.458	.5807392	3.342163
Black	1.110443	.1142772	1.02	0.309	.9076069	1.358609
Latino	1.676834	.1818305	4.77	0.000	1.355777	2.073918
Other	.5378856	.1941426	-1.72	0.086	.2651307	1.091239
AsianMale	.3983724	.2646617	-1.39	0.166	.1083396	1.464844
BlackMale	1.134735	.1842707	0.78	0.436	.8254045	1.559999
LatinoMale	1.015095	.1523652	0.10	0.920	.7563826	1.362296
OtherMale	2.445773	1.256981	1.74	0.082	.893201	6.697041
3						
tenure	1.012133	.0052534	2.32	0.020	1.001889	1.022482
exper	1.09398	.0080712	12.17	0.000	1.078274	1.109914
male	2.198702	.1877415	9.23	0.000	1.859878	2.599251
age	.9547332	.0060958	-7.26	0.000	.9428601	.9667559
pyacctrati~2	.9404754	.0512331	-1.13	0.260	.8452352	1.046447
pyperstudlep	.9942594	.0045998	-1.24	0.213	.9852848	1.003316
pyperstude~s	1.010418	.0031959	3.28	0.001	1.004174	1.016701
pyperstudw~e	1.014301	.0086166	1.67	0.095	.9975523	1.03133
pyperstudb~k	.9989611	.0098204	-0.11	0.916	.9798976	1.018395
pyperstudl~o	1.000851	.0091318	0.09	0.926	.9831121	1.01891
pyperstudm~e	1.003481	.0055799	0.62	0.532	.9926039	1.014477
Asian	1.876306	.9699802	1.22	0.223	.6811838	5.168245
Black	.7744033	.1139139	-1.74	0.082	.5804387	1.033185
Latino	1.601156	.2174294	3.47	0.001	1.227	2.089406
Other	1.068667	.3966447	0.18	0.858	.5163116	2.211938
AsianMale	.1919584	.1770332	-1.79	0.074	.0314911	1.170108
BlackMale	1.312724	.2791701	1.28	0.201	.8652725	1.991563
LatinoMale	.8174659	.1483836	-1.11	0.267	.5727465	1.166747
OtherMale	.6569705	.396064	-0.70	0.486	.2015519	2.141435

-----> pyschooltype = S

Iteration 0: log likelihood = -7786.4709  
 Iteration 1: log likelihood = -7472.4486  
 Iteration 2: log likelihood = -7459.2749  
 Iteration 3: log likelihood = -7458.9725  
 Iteration 4: log likelihood = -7458.9069  
 Iteration 5: log likelihood = -7458.8916  
 Iteration 6: log likelihood = -7458.8884  
 Iteration 7: log likelihood = -7458.8879  
 Iteration 8: log likelihood = -7458.8878  
 Iteration 9: log likelihood = -7458.8878

Multinomial logistic regression Number of obs = 8194

Log likelihood = -7458.8878

LR chi2(38) = 655.17  
 Prob > chi2 = 0.0000  
 Pseudo R2 = 0.0421

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.008628	.0037563	2.31	0.021	1.001293	1.016017
	exper	1.011773	.0047446	2.50	0.013	1.002516	1.021115
	male	1.122131	.0702439	1.84	0.066	.9925662	1.268609
	age	.9781214	.0038857	-5.57	0.000	.9705352	.9857668
pyacctrati~2		.9412025	.0373802	-1.53	0.127	.8707176	1.017393
pyperstudlep		1.018651	.0045833	4.11	0.000	1.009708	1.027674
pyperstude~s		.9882909	.0021696	-5.37	0.000	.9840477	.9925523
pyperstudw~e		.9989846	.0055172	-0.18	0.854	.9882295	1.009857
pyperstudb~k		1.004942	.0063387	0.78	0.434	.992595	1.017443
pyperstudl~o		1.00463	.005961	0.78	0.436	.9930145	1.016382
pyperstudm~e		1.000655	.0036205	0.18	0.856	.9935839	1.007776
	Asian	1.427116	.5524701	0.92	0.358	.6682504	3.04775
	Black	1.460998	.1470292	3.77	0.000	1.199466	1.779555
	Latino	1.829978	.1862732	5.94	0.000	1.499002	2.234033
	Other	.9361736	.317262	-0.19	0.846	.4818266	1.818955
	AsianMale	.5180806	.3244372	-1.05	0.294	.1518274	1.767846
	BlackMale	1.655443	.2376855	3.51	0.000	1.249396	2.193452
	LatinoMale	1.091524	.1436443	0.67	0.506	.8433645	1.412704
	OtherMale	.7488541	.367534	-0.59	0.556	.2861749	1.959579
3							
	tenure	.9961002	.004668	-0.83	0.404	.986993	1.005291
	exper	1.078042	.0079989	10.13	0.000	1.062478	1.093834
	male	1.967379	.1784112	7.46	0.000	1.647013	2.35006
	age	.9578994	.0066157	-6.23	0.000	.9450203	.9709541
pyacctrati~2		1.095614	.0629365	1.59	0.112	.9789518	1.22618
pyperstudlep		.9945335	.0073874	-0.74	0.461	.9801594	1.009118
pyperstude~s		1.013772	.0030345	4.57	0.000	1.007842	1.019737
pyperstudw~e		1.023383	.0098815	2.39	0.017	1.004197	1.042934
pyperstudb~k		1.013307	.0109893	1.22	0.223	.9919958	1.035076
pyperstudl~o		1.010921	.0103227	1.06	0.287	.9908895	1.031357
pyperstudm~e		.9892018	.0056942	-1.89	0.059	.978104	1.000425
	Asian	.5193551	.539249	-0.63	0.528	.0678668	3.974396
	Black	.9335803	.1825114	-0.35	0.725	.6364217	1.369488
	Latino	1.635219	.2742183	2.93	0.003	1.177155	2.271529
	Other	.3097375	.3166559	-1.15	0.252	.0417618	2.297249
	AsianMale	1.85e-06	.0011427	-0.02	0.983	0	.
	BlackMale	1.620367	.3976885	1.97	0.049	1.001618	2.621349
	LatinoMale	.8681401	.174543	-0.70	0.482	.5853995	1.287441
	OtherMale	5.258697	5.687797	1.53	0.125	.6312773	43.80626

## 2010-11

```
mlogit admin tenure exper male age pyacctrating2 pyperstudlep pyperstudecodis
pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian Black Latino Other
AsianMale Black
> Male LatinoMale OtherMale if validcert2011a==1, rrr iter(20)
```

```
Iteration 0: log likelihood = -28183.668
Iteration 1: log likelihood = -26997.724
Iteration 2: log likelihood = -26978.087
Iteration 3: log likelihood = -26978.062
Iteration 4: log likelihood = -26978.062
```

Multinomial logistic regression

Number of obs = 26437  
 LR chi2(38) = 2411.21

Log likelihood = -26978.062      Prob > chi2 = 0.0000  
Pseudo R2 = 0.0428

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.013975	.0024487	5.75	0.000	1.009187	1.018785
	exper	1.011743	.0029668	3.98	0.000	1.005945	1.017575
	male	1.221507	.0480661	5.08	0.000	1.130841	1.319443
	age	.9769446	.0022749	-10.02	0.000	.9724959	.9814136
pyacctrati~2		.9525619	.0206225	-2.24	0.025	.912988	.9938511
pyperstudlep		1.003834	.00114	3.37	0.001	1.001603	1.006071
pyperstude~s		.9933524	.0012286	-5.39	0.000	.9909473	.9957633
pyperstudw~e		.9934118	.0028598	-2.30	0.022	.9878224	.9990328
pyperstudb~k		.9949405	.0033273	-1.52	0.129	.9884404	1.001483
pyperstudl~o		.9974843	.0030747	-0.82	0.414	.9914762	1.003529
pyperstudm~e		1.006881	.0023209	2.98	0.003	1.002343	1.011441
	Asian	.8956567	.1881489	-0.52	0.600	.5933777	1.351923
	Black	1.162664	.064609	2.71	0.007	1.042685	1.296449
	Latino	1.41118	.0716152	6.79	0.000	1.277572	1.558761
	Other	.6503471	.1151514	-2.43	0.015	.4596532	.9201532
	AsianMale	1.926196	.7132152	1.77	0.077	.9322338	3.979935
	BlackMale	1.535417	.1414034	4.66	0.000	1.281845	1.839151
	LatinoMale	1.213712	.0933049	2.52	0.012	1.043949	1.411083
	OtherMale	1.809144	.527224	2.03	0.042	1.021915	3.202815
3							
	tenure	1.017016	.0024032	7.14	0.000	1.012317	1.021737
	exper	1.066657	.003502	19.65	0.000	1.059815	1.073543
	male	1.38212	.0573586	7.80	0.000	1.274149	1.499239
	age	.9721245	.0027279	-10.07	0.000	.9667926	.9774858
pyacctrati~2		1.264224	.0303091	9.78	0.000	1.206194	1.325047
pyperstudlep		1.009	.001237	7.31	0.000	1.006579	1.011427
pyperstude~s		1.017108	.0013611	12.68	0.000	1.014444	1.019779
pyperstudw~e		1.005195	.003371	1.55	0.122	.99861	1.011824
pyperstudb~k		.9908475	.0038607	-2.36	0.018	.9833094	.9984433
pyperstudl~o		.989712	.0035334	-2.90	0.004	.9828109	.9966616
pyperstudm~e		.9969762	.0027415	-1.10	0.271	.9916174	1.002364
	Asian	.7719864	.1961537	-1.02	0.308	.4691678	1.270255
	Black	.8552858	.0569496	-2.35	0.019	.7506433	.9745159
	Latino	1.11982	.0640459	1.98	0.048	1.001072	1.252654
	Other	.768359	.1446263	-1.40	0.162	.531306	1.111178
	AsianMale	1.285531	.62195	0.52	0.604	.4980407	3.318185
	BlackMale	1.311301	.1482438	2.40	0.017	1.050687	1.636559
	LatinoMale	1.064569	.0933018	0.71	0.475	.8965447	1.264082
	OtherMale	1.527682	.484083	1.34	0.181	.8209361	2.842869

. mlogtest, all

\*\*\*\* Likelihood-ratio tests for independent variables (N=26437)

Ho: All coefficients associated with given variable(s) are 0.

	chi2	df	P>chi2
tenure	60.894	2	0.000
exper	420.359	2	0.000
male	66.773	2	0.000
age	158.202	2	0.000
pyacctrati~2	136.314	2	0.000
pyperstudlep	53.672	2	0.000
pyperstude~s	281.120	2	0.000
pyperstudw~e	11.930	2	0.003

pyperstudb~k		6.046	2	0.049
pyperstudl~o		8.310	2	0.016
pyperstudm~e		14.377	2	0.001
Asian		1.120	2	0.571
Black		19.263	2	0.000
Latino		46.313	2	0.000
Other		6.749	2	0.034
AsianMale		3.174	2	0.205
BlackMale		22.240	2	0.000
LatinoMale		6.511	2	0.039
OtherMale		4.648	2	0.098

\*\*\*\* Wald tests for independent variables (N=26437)

Ho: All coefficients associated with given variable(s) are 0.

		chi2	df	P>chi2
tenure		60.784	2	0.000
exper		392.747	2	0.000
male		66.660	2	0.000
age		155.423	2	0.000
pyacctrati~2		134.890	2	0.000
pyperstudlep		53.579	2	0.000
pyperstude~s		274.889	2	0.000
pyperstudw~e		11.946	2	0.003
pyperstudb~k		6.060	2	0.048
pyperstudl~o		8.468	2	0.014
pyperstudm~e		14.433	2	0.001
Asian		1.093	2	0.579
Black		19.189	2	0.000
Latino		46.484	2	0.000
Other		6.547	2	0.038
AsianMale		3.154	2	0.207
BlackMale		22.059	2	0.000
LatinoMale		6.508	2	0.039
OtherMale		4.608	2	0.100

\*\*\*\* Hausman tests of IIA assumption (N=26437)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		chi2	df	P>chi2	evidence
2		-91.876	19	---	---
3		5.198	19	0.999	for Ho

Note: If chi2<0, the estimated model does not meet asymptotic assumptions of the test.

\*\*\*\* Small-Hsiao tests of IIA assumption (N=26437)

Ho: Odds(Outcome-J vs Outcome-K) are independent of other alternatives.

Omitted		lnL(full)	lnL(omit)	chi2	df	P>chi2	evidence
2		-5526.373	-5515.467	21.813	20	0.351	for Ho
3		-6580.084	-6573.874	12.418	20	0.901	for Ho

\*\*\*\* Wald tests for combining alternatives (N=26437)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be combined).

Alternatives tested		chi2	df	P>chi2
2-	3	1406.910	19	0.000
2-	1	493.954	19	0.000
3-	1	1531.835	19	0.000

\*\*\*\* LR tests for combining alternatives (N=26437)

Ho: All coefficients except intercepts associated with a given pair of alternatives are 0 (i.e., alternatives can be collapsed).

Alternatives tested		chi2	df	P>chi2
2-	3	1540.008	19	0.000
2-	1	508.739	19	0.000
3-	1	1687.182	19	0.000

by region:

```
. bysort region: mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstuddecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino Other
```

```
> AsianMale BlackMale LatinoMale OtherMale if validcert2011a==1, rrr iter(20)
```

```
-----> region = 1
```

note: OtherMale omitted because of collinearity

```
Iteration 0: log likelihood = -2015.1404
Iteration 1: log likelihood = -1917.6324
Iteration 2: log likelihood = -1914.8399
Iteration 3: log likelihood = -1914.5066
Iteration 4: log likelihood = -1914.4519
Iteration 5: log likelihood = -1914.4462
Iteration 6: log likelihood = -1914.445
Iteration 7: log likelihood = -1914.4447
Iteration 8: log likelihood = -1914.4447
```

Multinomial logistic regression	Number of obs	=	1864
	LR chi2(36)	=	201.39
	Prob > chi2	=	0.0000
Log likelihood = -1914.4447	Pseudo R2	=	0.0500

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.020146	.0091006	2.24	0.025	1.002464	1.03814
	exper	1.003958	.0121449	0.33	0.744	.9804345	1.028046
	male	.891176	.3072857	-0.33	0.738	.4533789	1.751724
	age	.9673449	.0098378	-3.26	0.001	.9482541	.98682
pyacctrati~2		.9153526	.0744588	-1.09	0.277	.7804552	1.073566
pyperstudlep		1.002514	.0026963	0.93	0.351	.9972429	1.007812
pyperstude~s		1.003255	.0053537	0.61	0.543	.9928162	1.013803
pyperstudw~e		.9795813	.0546204	-0.37	0.711	.8781695	1.092704
pyperstudb~k		.8428201	.1149265	-1.25	0.210	.6451571	1.101043
pyperstudl~o		.9754765	.0460198	-0.53	0.599	.8893238	1.069975
pyperstudm~e		1.006638	.0101387	0.66	0.511	.9869618	1.026707
	Asian	.6738951	.6053972	-0.44	0.660	.115854	3.919889
	Black	.5896797	.699115	-0.45	0.656	.0577358	6.022647
	Latino	1.381213	.3042649	1.47	0.143	.8969149	2.127014
	Other	.3199133	.3577743	-1.02	0.308	.035735	2.863989
	AsianMale	1.13e-06	.002047	-0.01	0.994	0	



BlackMale		16.99732	28.20931	1.71	0.088	.6572014	439.6048
LatinoMale		1.764366	.6458541	1.55	0.121	.8609984	3.615554
OtherMale		(omitted)					

-----

3							
tenure		1.018595	.009236	2.03	0.042	1.000653	1.036859
exper		1.095329	.0170562	5.85	0.000	1.062405	1.129274
male		.6834323	.243452	-1.07	0.285	.3400015	1.373758
age		.9330392	.0131657	-4.91	0.000	.9075885	.9592035
pyacctrati~2		1.126695	.1054051	1.28	0.202	.9379386	1.353438
pyperstudlep		1.015037	.003055	4.96	0.000	1.009067	1.021042
pyperstude~s		.9988244	.0060943	-0.19	0.847	.9869509	1.010841
pyperstudw~e		1.026134	.0641926	0.41	0.680	.907726	1.159987
pyperstudb~k		.8727653	.125938	-0.94	0.346	.6577655	1.158041
pyperstudl~o		1.007458	.0529027	0.14	0.887	.9089282	1.116669
pyperstudm~e		.9910214	.0120778	-0.74	0.459	.9676298	1.014978
Asian		8.88e-07	.0005819	-0.02	0.983	0	.
Black		6.12e-07	.0005279	-0.02	0.987	0	.
Latino		.7234146	.1556425	-1.50	0.132	.4745165	1.102867
Other		.3766887	.335218	-1.10	0.273	.0658404	2.155128
AsianMale		4.165854	9711.84	0.00	1.000	0	.
BlackMale		14.21118	16383.61	0.00	0.998	0	.
LatinoMale		2.021089	.7788516	1.83	0.068	.9496519	4.301366
OtherMale		(omitted)					

-----

-----> region = 2

```

Iteration 0: log likelihood = -723.72095
Iteration 1: log likelihood = -682.9685
Iteration 2: log likelihood = -680.89563
Iteration 3: log likelihood = -680.51304
Iteration 4: log likelihood = -680.43539
Iteration 5: log likelihood = -680.41861
Iteration 6: log likelihood = -680.41506
Iteration 7: log likelihood = -680.41425
Iteration 8: log likelihood = -680.41412
Iteration 9: log likelihood = -680.41409
Iteration 10: log likelihood = -680.41408

```

Multinomial logistic regression	Number of obs	=	687
	LR chi2(38)	=	86.61
	Prob > chi2	=	0.0000
Log likelihood = -680.41408	Pseudo R2	=	0.0598

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.015154	.0149263	1.02	0.306	.9863167 1.044835
exper		.9884042	.018027	-0.64	0.522	.953696 1.024376
male		1.511341	.4323451	1.44	0.149	.862704 2.647667
age		.9864324	.0141576	-0.95	0.341	.9590707 1.014575
pyacctrati~2		1.109991	.1645611	0.70	0.482	.8300897 1.484273
pyperstudlep		1.006516	.0203871	0.32	0.748	.9673406 1.047278
pyperstude~s		1.000099	.0074636	0.01	0.989	.9855772 1.014835
pyperstudw~e		.9520433	.054571	-0.86	0.391	.8508752 1.06524
pyperstudb~k		.9204662	.0640249	-1.19	0.233	.8031576 1.054909
pyperstudl~o		.9526072	.052323	-0.88	0.377	.8553832 1.060882
pyperstudm~e		1.006782	.0137421	0.50	0.620	.9802056 1.03408
Asian		1.070879	1.341496	0.05	0.956	.0919254 12.47514
Black		1.302671	1.184457	0.29	0.771	.2192184 7.740912
Latino		1.174086	.3035529	0.62	0.535	.7073378 1.948826
Other		3.21e-07	.0008322	-0.01	0.995	0 .

AsianMale		9.99e-08	.0002878	-0.01	0.996	0	.
BlackMale		1.39e-07	.000233	-0.01	0.992	0	.
LatinoMale		1.231947	.491973	0.52	0.601	.5632069	2.694734
OtherMale		1335758	3.47e+09	0.01	0.996	0	.
-----							
3							
tenure		1.025626	.0144005	1.80	0.072	.9977864	1.054242
exper		.9918862	.018201	-0.44	0.657	.9568468	1.028209
male		.7940731	.2575351	-0.71	0.477	.4205303	1.499421
age		1.012195	.0153651	0.80	0.425	.9825236	1.042763
pyacctrati~2		1.608683	.2469924	3.10	0.002	1.190638	2.173509
pyperstudlep		1.011703	.0196812	0.60	0.550	.9738545	1.051022
pyperstude~s		1.016322	.0085305	1.93	0.054	.9997397	1.03318
pyperstudw~e		1.20303	.0927857	2.40	0.017	1.034252	1.399352
pyperstudb~k		1.108709	.1041653	1.10	0.272	.9222432	1.332875
pyperstudl~o		1.183169	.0874933	2.27	0.023	1.023533	1.367703
pyperstudm~e		.9638525	.016879	-2.10	0.036	.9313316	.9975089
Asian		3.90e-07	.000718	-0.01	0.994	0	.
Black		2.568481	2.161585	1.12	0.262	.4935386	13.36692
Latino		1.138494	.2973994	0.50	0.620	.6823046	1.899691
Other		5.689469	6.723051	1.47	0.141	.561353	57.66436
AsianMale		.6774346	2312.841	-0.00	1.000	0	.
BlackMale		.5561416	.8115523	-0.40	0.688	.0318468	9.711926
LatinoMale		1.912238	.835313	1.48	0.138	.8122988	4.50161
OtherMale		7.05e-08	.0001154	-0.01	0.992	0	.
-----							

-----> region = 3

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -318.07392  
Iteration 1: log likelihood = -281.80799  
Iteration 2: log likelihood = -281.14572  
Iteration 3: log likelihood = -281.09228  
Iteration 4: log likelihood = -281.08328  
Iteration 5: log likelihood = -281.0812  
Iteration 6: log likelihood = -281.08071  
Iteration 7: log likelihood = -281.0806  
Iteration 8: log likelihood = -281.08058

Multinomial logistic regression	Number of obs	=	300
	LR chi2(32)	=	73.99
	Prob > chi2	=	0.0000
Log likelihood = -281.08058	Pseudo R2	=	0.1163

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.009993	.0264664	0.38	0.704	.959429 1.063221
exper		.9806187	.0301225	-0.64	0.524	.9233219 1.041471
male		1.654281	.6379608	1.31	0.192	.7768732 3.522643
age		.9772637	.0243109	-0.92	0.355	.9307582 1.026093
pyacctrati~2		1.416101	.3761017	1.31	0.190	.8414412 2.383222
pyperstudlep		1.000415	.0335468	0.01	0.990	.9367784 1.068374
pyperstude~s		1.030582	.0228959	1.36	0.175	.9866694 1.076448
pyperstudw~e		1.02253	.0611092	0.37	0.709	.9095069 1.149599
pyperstudb~k		1.039297	.0652137	0.61	0.539	.9190272 1.175305
pyperstudl~o		1.017377	.066827	0.26	0.793	.8944789 1.15716
pyperstudm~e		1.029465	.0250916	1.19	0.233	.9814425 1.079837
Asian		(omitted)				
Black		1.234401	1.03518	0.25	0.802	.2385781 6.386777

Latino		2.805277	1.904184	1.52	0.129	.7416372	10.61109
Other		1.01e+07	1.58e+10	0.01	0.992	0	.
AsianMale		(omitted)					
BlackMale		.75707	716.2082	-0.00	1.000	0	.
LatinoMale		.1251813	.1680179	-1.55	0.122	.009017	1.737876
OtherMale		(omitted)					
-----							
3							
tenure		1.002152	.0204203	0.11	0.916	.962918	1.042985
exper		1.012573	.0253121	0.50	0.617	.9641584	1.06342
male		3.095214	1.099571	3.18	0.001	1.542775	6.209819
age		1.026133	.0224768	1.18	0.239	.9830121	1.071146
pyacctrati~2		1.851931	.4544068	2.51	0.012	1.144897	2.995597
pyperstudlep		1.029966	.0316416	0.96	0.337	.96978	1.093888
pyperstude~s		1.064639	.0217492	3.07	0.002	1.022853	1.108131
pyperstudw~e		1.097123	.0626476	1.62	0.105	.9809577	1.227045
pyperstudb~k		1.058025	.0638587	0.93	0.350	.9399835	1.190889
pyperstudl~o		1.059046	.0659314	0.92	0.357	.9373955	1.196484
pyperstudm~e		1.001994	.0307557	0.06	0.948	.9434918	1.064125
Asian		(omitted)					
Black		.9162336	.7961254	-0.10	0.920	.1668737	5.030655
Latino		2.351458	1.647231	1.22	0.222	.5957414	9.281467
Other		.9549989	2263.043	-0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		802380.1	4.55e+08	0.02	0.981	0	.
LatinoMale		.3406573	.3658765	-1.00	0.316	.041505	2.795984
OtherMale		(omitted)					
-----							

-----> region = 4

Iteration 0: log likelihood = -5795.2043  
Iteration 1: log likelihood = -5464.2029  
Iteration 2: log likelihood = -5447.9087  
Iteration 3: log likelihood = -5447.82  
Iteration 4: log likelihood = -5447.82

Multinomial logistic regression	Number of obs	=	5619
	LR chi2(38)	=	694.77
	Prob > chi2	=	0.0000
Log likelihood = -5447.82	Pseudo R2	=	0.0599

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.021134	.0054966	3.89	0.000	1.010417 1.031964
exper		1.035054	.0068907	5.18	0.000	1.021636 1.048648
male		1.190321	.1051233	1.97	0.049	1.001129 1.415266
age		.9668918	.0050512	-6.44	0.000	.9570421 .9768428
pyacctrati~2		.9631389	.0474558	-0.76	0.446	.8744772 1.06079
pyperstudlep		.9988466	.0025297	-0.46	0.649	.9939007 1.003817
pyperstude~s		.9992145	.0036178	-0.22	0.828	.9921489 1.00633
pyperstudw~e		.999317	.0045283	-0.15	0.880	.990481 1.008232
pyperstudb~k		.9982039	.0058253	-0.31	0.758	.9868516 1.009687
pyperstudl~o		1.000043	.0055958	0.01	0.994	.9891357 1.011071
pyperstudm~e		1.008166	.0055756	1.47	0.141	.9972967 1.019153
Asian		.7216554	.254006	-0.93	0.354	.362015 1.438577
Black		.8842385	.0857687	-1.27	0.205	.7311478 1.069384
Latino		1.551507	.1803608	3.78	0.000	1.235385 1.94852
Other		.7965675	.2620077	-0.69	0.489	.4180658 1.517751
AsianMale		1.848305	1.226506	0.93	0.355	.5034187 6.78606
BlackMale		1.502385	.2388145	2.56	0.010	1.100213 2.051567
LatinoMale		1.095986	.2131753	0.47	0.637	.7485859 1.604606

OtherMale		1.554867	1.014823	0.68	0.499	.4326486	5.587936
-----							
3							
tenure		1.031775	.0060952	5.30	0.000	1.019898	1.043791
exper		1.113946	.0100279	11.99	0.000	1.094464	1.133775
male		1.301196	.1407251	2.43	0.015	1.052652	1.608424
age		.9402152	.0073055	-7.93	0.000	.9260052	.9546434
pyacctrati~2		1.314962	.0819187	4.40	0.000	1.163819	1.485733
pyperstudlep		1.017059	.0031618	5.44	0.000	1.010881	1.023275
pyperstude~s		1.023922	.0044816	5.40	0.000	1.015176	1.032744
pyperstudw~e		1.010437	.0057393	1.83	0.068	.99925	1.021748
pyperstudb~k		.9974549	.0073174	-0.35	0.728	.9832157	1.0119
pyperstudl~o		.9818275	.0067463	-2.67	0.008	.9686936	.9951394
pyperstudm~e		.9968244	.0071256	-0.44	0.656	.982956	1.010889
Asian		.6208068	.2788541	-1.06	0.289	.257403	1.497267
Black		.5973451	.075894	-4.06	0.000	.4656701	.7662532
Latino		1.150173	.1650244	0.98	0.329	.8682296	1.523673
Other		.4427936	.2431127	-1.48	0.138	.1509577	1.298815
AsianMale		1.6547	1.557792	0.53	0.593	.261438	10.47297
BlackMale		1.459706	.3106825	1.78	0.076	.9618259	2.215309
LatinoMale		1.201475	.2971908	0.74	0.458	.7398877	1.951029
OtherMale		4.951913	4.217672	1.88	0.060	.9327877	26.28834
-----							

-----> region = 5

Iteration 0: log likelihood = -520.15513  
Iteration 1: log likelihood = -479.36809  
Iteration 2: log likelihood = -477.79323  
Iteration 3: log likelihood = -477.61452  
Iteration 4: log likelihood = -477.58759  
Iteration 5: log likelihood = -477.58109  
Iteration 6: log likelihood = -477.57972  
Iteration 7: log likelihood = -477.57943  
Iteration 8: log likelihood = -477.57937  
Iteration 9: log likelihood = -477.57935

Multinomial logistic regression	Number of obs	=	498
	LR chi2(38)	=	85.15
	Prob > chi2	=	0.0000
Log likelihood = -477.57935	Pseudo R2	=	0.0819

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		.9786617	.0174448	-1.21	0.226	.9450609 1.013457
exper		1.010826	.0224033	0.49	0.627	.9678568 1.055704
male		1.541697	.4280162	1.56	0.119	.8947109 2.656534
age		1.010138	.0182196	0.56	0.576	.9750521 1.046486
pyacctrati~2		.985902	.1711865	-0.08	0.935	.7015132 1.38558
pyperstudlep		1.016426	.0328283	0.50	0.614	.954078 1.082849
pyperstude~s		.9975246	.0115676	-0.21	0.831	.9751083 1.020456
pyperstudw~e		.9762965	.0490794	-0.48	0.633	.8846896 1.077389
pyperstudb~k		.9686978	.0503085	-0.61	0.540	.8749473 1.072494
pyperstudl~o		.9702786	.0616141	-0.48	0.635	.8567298 1.098877
pyperstudm~e		1.031343	.0278515	1.14	0.253	.978175 1.087402
Asian		3.297937	5.040134	0.78	0.435	.1649597 65.93363
Black		1.820383	.7297547	1.49	0.135	.829723 3.993855
Latino		.9400952	.8142841	-0.07	0.943	.1721412 5.134034
Other		7.312908	8.838561	1.65	0.100	.6843953 78.13998
AsianMale		.170866	767.3316	-0.00	1.000	0 .
BlackMale		2.638138	1.497543	1.71	0.087	.8671843 8.025711
LatinoMale		3.91e-07	.0005423	-0.01	0.992	0 .

OtherMale		.1865739	.3537472	-0.89	0.376	.004539	7.669077
-----							
3							
tenure		1.000687	.0180522	0.04	0.970	.9659235	1.036701
exper		1.060872	.0271136	2.31	0.021	1.009039	1.115367
male		2.207842	.6446478	2.71	0.007	1.245757	3.912935
age		.9822277	.0217921	-0.81	0.419	.9404314	1.025882
pyacctrati~2		1.54835	.2848582	2.38	0.017	1.079616	2.220595
pyperstudlep		.9819237	.0340828	-0.53	0.599	.9173442	1.051049
pyperstude~s		1.031173	.0126421	2.50	0.012	1.006691	1.056252
pyperstudw~e		.9539311	.0477954	-0.94	0.347	.8647065	1.052362
pyperstudb~k		.9243109	.0480379	-1.51	0.130	.8347949	1.023426
pyperstudl~o		.9499106	.0596655	-0.82	0.413	.8398801	1.074356
pyperstudm~e		1.015848	.0292178	0.55	0.585	.9601664	1.074759
Asian		2.71e-06	.0031483	-0.01	0.991	0	.
Black		1.961971	.9065827	1.46	0.145	.7931796	4.853038
Latino		.4976932	.5715694	-0.61	0.543	.0524095	4.726213
Other		3.663735	5.68076	0.84	0.402	.17543	76.51458
AsianMale		5.93e+12	1.72e+16	0.01	0.992	0	.
BlackMale		1.644677	1.062251	0.77	0.441	.4637816	5.832408
LatinoMale		5.68e-07	.0007122	-0.01	0.991	0	.
OtherMale		1.90e-07	.0002254	-0.01	0.990	0	.
-----							

-----> region = 6

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1079.9338  
Iteration 1: log likelihood = -1011.2463  
Iteration 2: log likelihood = -1009.263  
Iteration 3: log likelihood = -1009.0837  
Iteration 4: log likelihood = -1009.0508  
Iteration 5: log likelihood = -1009.0436  
Iteration 6: log likelihood = -1009.0419  
Iteration 7: log likelihood = -1009.0415  
Iteration 8: log likelihood = -1009.0415  
Iteration 9: log likelihood = -1009.0414

Multinomial logistic regression	Number of obs	=	1028
	LR chi2(36)	=	141.78
	Prob > chi2	=	0.0000
Log likelihood = -1009.0414	Pseudo R2	=	0.0656

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.01593	.0135077	1.19	0.235	.9897977	1.042753
	exper	1.010514	.0165289	0.64	0.523	.9786317	1.043435
	male	1.413242	.24848	1.97	0.049	1.001281	1.994698
	age	.9737434	.0124169	-2.09	0.037	.9497084	.9983867
pyacctrati~2		1.023102	.1207637	0.19	0.847	.811794	1.289413
pyperstudlep		1.013779	.011626	1.19	0.233	.9912468	1.036824
pyperstude~s		.9877199	.0068681	-1.78	0.076	.9743499	1.001273
pyperstudw~e		1.050597	.0318281	1.63	0.103	.9900308	1.114868
pyperstudb~k		1.033661	.0343791	1.00	0.320	.9684281	1.103287
pyperstudl~o		1.053787	.0325452	1.70	0.090	.9918922	1.119545
pyperstudm~e		1.005933	.0157571	0.38	0.706	.9755191	1.037296
	Asian	.563774	.658829	-0.49	0.624	.0570673	5.56959
	Black	2.362292	.886154	2.29	0.022	1.13247	4.927654
	Latino	.6457341	.2958594	-0.95	0.340	.2630597	1.585088
	Other	1.29e-06	.0015206	-0.01	0.991	0	.
	AsianMale	(omitted)					
	BlackMale	4.494518	3.437602	1.96	0.049	1.0038	20.12422

LatinoMale		1.417667	.9561951	0.52	0.605	.3779647	5.317371
OtherMale		1.30e+13	3.98e+16	0.01	0.992	0	.
-----							
3							
tenure		.9981919	.0125174	-0.14	0.885	.9739573	1.023029
exper		1.097503	.0194831	5.24	0.000	1.059974	1.136362
male		1.628723	.3058685	2.60	0.009	1.127183	2.353422
age		.9638009	.0146009	-2.43	0.015	.9356044	.9928473
pyacctrati~2		1.231503	.1535072	1.67	0.095	.9645684	1.572309
pyperstudlep		1.024422	.0126113	1.96	0.050	.9999998	1.04944
pyperstude~s		1.017782	.0074426	2.41	0.016	1.003299	1.032474
pyperstudw~e		1.015102	.0294859	0.52	0.606	.9589253	1.07457
pyperstudb~k		1.013647	.0317465	0.43	0.665	.9532959	1.077818
pyperstudl~o		.9948523	.0292943	-0.18	0.861	.9390619	1.053957
pyperstudm~e		.9912064	.0179126	-0.49	0.625	.9567129	1.026944
Asian		6.60e-07	.0007164	-0.01	0.990	0	.
Black		1.302673	.5706019	0.60	0.546	.5520632	3.073846
Latino		.5034606	.2891163	-1.20	0.232	.1633639	1.551583
Other		.9072934	1.157927	-0.08	0.939	.0743717	11.06847
AsianMale		(omitted)					
BlackMale		3.115629	2.687461	1.32	0.188	.5745412	16.89547
LatinoMale		1.857189	1.465605	0.78	0.433	.3954836	8.721347
OtherMale		.2497657	1226.807	-0.00	1.000	0	.
-----							

-----> region = 7

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1198.5867  
Iteration 1: log likelihood = -1122.3105  
Iteration 2: log likelihood = -1120.3388  
Iteration 3: log likelihood = -1120.2125  
Iteration 4: log likelihood = -1120.1926  
Iteration 5: log likelihood = -1120.1877  
Iteration 6: log likelihood = -1120.1866  
Iteration 7: log likelihood = -1120.1863  
Iteration 8: log likelihood = -1120.1863  
Iteration 9: log likelihood = -1120.1863

Multinomial logistic regression	Number of obs	=	1157
	LR chi2(36)	=	156.80
	Prob > chi2	=	0.0000
Log likelihood = -1120.1863	Pseudo R2	=	0.0654

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9918952	.0116031	-0.70	0.487	.9694123	1.0149
	exper	1.028108	.0142044	2.01	0.045	1.000642	1.056329
	male	1.777178	.317333	3.22	0.001	1.252392	2.521863
	age	.9672815	.01139	-2.83	0.005	.9452131	.9898652
pyacctrati~2		.9029291	.1075307	-0.86	0.391	.7149625	1.140313
pyperstudlep		1.011759	.0125208	0.94	0.345	.9875138	1.036599
pyperstude~s		.9989738	.0081349	-0.13	0.900	.9831563	1.015046
pyperstudw~e		.963965	.0516002	-0.69	0.493	.867955	1.070595
pyperstudb~k		.9627118	.0536585	-0.68	0.495	.863084	1.07384
pyperstudl~o		.9707928	.0528268	-0.54	0.586	.8725844	1.080055
pyperstudm~e		.9822262	.0179691	-0.98	0.327	.9476313	1.018084
	Asian	7048231	2.45e+10	0.00	0.996	0	.
	Black	2.053558	.6168911	2.40	0.017	1.139738	3.70006
	Latino	2.584147	1.906747	1.29	0.198	.6084811	10.97457
	Other	2.332855	1.93338	1.02	0.307	.4596686	11.83943
	AsianMale	(omitted)					

BlackMale		1.512604	.7665175	0.82	0.414	.5602421	4.083896
LatinoMale		3.555545	4.688997	0.96	0.336	.2681359	47.14736
OtherMale		.4897026	.6143233	-0.57	0.569	.0418904	5.724662
-----							
3							
tenure		.9960002	.0104412	-0.38	0.702	.9757447	1.016676
exper		1.045588	.0136579	3.41	0.001	1.019159	1.072702
male		2.111807	.35441	4.45	0.000	1.519857	2.934306
age		.983243	.0112467	-1.48	0.140	.9614452	1.005535
pyacctrati~2		1.046022	.1184948	0.40	0.691	.8377523	1.306069
pyperstudlep		1.044071	.0136734	3.29	0.001	1.017613	1.071217
pyperstude~s		1.035258	.0078148	4.59	0.000	1.020054	1.050689
pyperstudw~e		.885591	.0432312	-2.49	0.013	.8047866	.9745085
pyperstudb~k		.8734883	.0444642	-2.66	0.008	.7905464	.9651322
pyperstudl~o		.8440166	.0420997	-3.40	0.001	.7654078	.9306986
pyperstudm~e		.9846194	.0186668	-0.82	0.414	.9487045	1.021894
Asian		.0844554	497.6888	-0.00	1.000	0	.
Black		1.073316	.3928943	0.19	0.847	.5237693	2.199457
Latino		3.373883	2.510588	1.63	0.102	.7847485	14.5054
Other		1.457563	1.343465	0.41	0.683	.2393634	8.87558
AsianMale		(omitted)					
BlackMale		1.850936	1.044994	1.09	0.275	.6121014	5.597052
LatinoMale		4.461306	5.913089	1.13	0.259	.3320974	59.93198
OtherMale		5.38e-07	.0003919	-0.02	0.984	0	.
-----							

-----> region = 8

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -387.37646  
Iteration 1: log likelihood = -346.11158  
Iteration 2: log likelihood = -344.39083  
Iteration 3: log likelihood = -344.03834  
Iteration 4: log likelihood = -343.96061  
Iteration 5: log likelihood = -343.94435  
Iteration 6: log likelihood = -343.94164  
Iteration 7: log likelihood = -343.941  
Iteration 8: log likelihood = -343.94086  
Iteration 9: log likelihood = -343.94083

Multinomial logistic regression	Number of obs	=	365
	LR chi2(36)	=	86.87
	Prob > chi2	=	0.0000
Log likelihood = -343.94083	Pseudo R2	=	0.1121

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1		(base outcome)				
-----						
2						
tenure		1.03433	.0236542	1.48	0.140	.9889922 1.081746
exper		.9869625	.0286525	-0.45	0.651	.9323724 1.044749
male		3.144335	1.029613	3.50	0.000	1.655018 5.973858
age		.9798686	.0236492	-0.84	0.399	.9345962 1.027334
pyacctrati~2		.8342703	.185369	-0.82	0.415	.539729 1.289549
pyperstudlep		1.055469	.029684	1.92	0.055	.9988638 1.115282
pyperstude~s		.983754	.0140992	-1.14	0.253	.9565045 1.01178
pyperstudw~e		1.060559	.159762	0.39	0.696	.789424 1.424818
pyperstudb~k		1.085372	.1635642	0.54	0.587	.8077991 1.458322
pyperstudl~o		1.063016	.1583167	0.41	0.682	.7939056 1.423346
pyperstudm~e		1.062774	.050055	1.29	0.196	.96906 1.165551
Asian		.7872784	5314.336	-0.00	1.000	0 .
Black		3.486512	2.150229	2.03	0.043	1.040953 11.67753
Latino		1.22e-06	.0024796	-0.01	0.995	0 .
Other		7.94e-07	.0013059	-0.01	0.993	0 .

AsianMale		(omitted)					
BlackMale		937823.1	8.73e+08	0.01	0.988	0	.
LatinoMale		1858184	1.31e+10	0.00	0.998	0	.
OtherMale		1435322	9.97e+09	0.00	0.998	0	.
-----							
3							
tenure		1.021794	.0197464	1.12	0.265	.9838155	1.061238
exper		1.007429	.0262718	0.28	0.777	.9572314	1.06026
male		3.03698	.9047916	3.73	0.000	1.693745	5.445478
age		.9893763	.0223713	-0.47	0.637	.9464868	1.034209
pyacctrati~2		1.207368	.2445204	0.93	0.352	.8118038	1.795676
pyperstudlep		1.055503	.029572	1.93	0.054	.9991054	1.115084
pyperstude~s		1.0136	.0126087	1.09	0.278	.989186	1.038616
pyperstudw~e		1.119715	.1397263	0.91	0.365	.8767748	1.429969
pyperstudb~k		1.09619	.1372518	0.73	0.463	.8576468	1.401081
pyperstudl~o		1.081584	.1335549	0.64	0.525	.8490892	1.37774
pyperstudm~e		1.075799	.0459492	1.71	0.087	.9894068	1.169735
Asian		4997671	1.96e+10	0.00	0.997	0	.
Black		3.790168	2.338551	2.16	0.031	1.131002	12.70146
Latino		6.5679	8.76985	1.41	0.159	.4795696	89.95005
Other		5.427876	6.503078	1.41	0.158	.5185789	56.81264
AsianMale		(omitted)					
BlackMale		1014612	9.45e+08	0.01	0.988	0	.
LatinoMale		2284069	8.96e+09	0.00	0.997	0	.
OtherMale		4305732	1.69e+10	0.00	0.997	0	.
-----							

-----> region = 9

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
Iteration 0: log likelihood = -251.07424  
Iteration 1: log likelihood = -221.21678  
Iteration 2: log likelihood = -219.67315  
Iteration 3: log likelihood = -219.45075  
Iteration 4: log likelihood = -219.40852  
Iteration 5: log likelihood = -219.39813  
Iteration 6: log likelihood = -219.39602  
Iteration 7: log likelihood = -219.39572  
Iteration 8: log likelihood = -219.39568  
Iteration 9: log likelihood = -219.39568

Multinomial logistic regression	Number of obs	=	234
	LR chi2(34)	=	63.36
	Prob > chi2	=	0.0016
Log likelihood = -219.39568	Pseudo R2	=	0.1262

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9762298	.0300471	-0.78	0.434	.9190796	1.036934
	exper	.9794326	.0367138	-0.55	0.579	.9100546	1.0541
	male	.5046189	.2133373	-1.62	0.106	.2203443	1.155647
	age	.9830165	.0296092	-0.57	0.570	.9266633	1.042797
pyacctrati~2		.6538348	.2061691	-1.35	0.178	.3524234	1.213029
pyperstudlep		1.1288	.0723933	1.89	0.059	.9954669	1.279991
pyperstude~s		1.001413	.0194862	0.07	0.942	.9639402	1.040343
pyperstudw~e		.7776777	.0920564	-2.12	0.034	.6166518	.9807522
pyperstudb~k		.7908087	.100893	-1.84	0.066	.6158477	1.015476
pyperstudl~o		.7584195	.093313	-2.25	0.025	.5959105	.965246
pyperstudm~e		1.045618	.0456376	1.02	0.307	.9598891	1.139003
	Asian	(omitted)					
	Black	1.599382	2.176078	0.35	0.730	.1111296	23.01839



Latino		2.10e+07	1.03e+11	0.00	0.997	0	.
Other		5.67e+07	2.79e+11	0.00	0.997	0	.
AsianMale		(omitted)					
BlackMale		8744676	1.27e+10	0.01	0.991	0	.
LatinoMale		7.19e-08	.0004061	-0.00	0.998	0	.
OtherMale		.1228222	629.0955	-0.00	1.000	0	.
-----							
3							
tenure		1.033023	.0233881	1.44	0.151	.9881855	1.079895
exper		1.024846	.0344713	0.73	0.466	.9594623	1.094685
male		1.281625	.4441597	0.72	0.474	.6497835	2.527861
age		.9656107	.0281398	-1.20	0.230	.9120033	1.022369
pyacctrati~2		.8747018	.2224737	-0.53	0.599	.5313289	1.439981
pyperstudlep		1.136727	.0636944	2.29	0.022	1.018499	1.268679
pyperstude~s		1.015421	.0159877	0.97	0.331	.9845644	1.047245
pyperstudw~e		.9660475	.0956133	-0.35	0.727	.7957047	1.172857
pyperstudb~k		.9377149	.1027093	-0.59	0.557	.7565497	1.162262
pyperstudl~o		.9296322	.09437	-0.72	0.472	.7619087	1.134278
pyperstudm~e		1.023519	.037161	0.64	0.522	.9532156	1.099007
Asian		(omitted)					
Black		2.789388	3.772562	0.76	0.448	.1969178	39.51235
Latino		1.588042	11222.98	0.00	1.000	0	.
Other		.8105252	5728.13	-0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		863318	1.25e+09	0.01	0.992	0	.
LatinoMale		2897641	2.10e+10	0.00	0.998	0	.
OtherMale		3586636	2.59e+10	0.00	0.998	0	.
-----							

-----> region = 10

Iteration 0: log likelihood = -4323.142  
Iteration 1: log likelihood = -4073.3132  
Iteration 2: log likelihood = -4067.1581  
Iteration 3: log likelihood = -4067.1307  
Iteration 4: log likelihood = -4067.1307

Multinomial logistic regression	Number of obs	=	4075
	LR chi2(38)	=	512.02
	Prob > chi2	=	0.0000
Log likelihood = -4067.1307	Pseudo R2	=	0.0592

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.017879	.0066453	2.71	0.007	1.004938 1.030987
exper		1.033385	.0080863	4.20	0.000	1.017657 1.049356
male		1.10436	.1120491	0.98	0.328	.9052058 1.347331
age		.9627714	.0059215	-6.17	0.000	.9512352 .9744476
pyacctrati~2		.9197099	.0521889	-1.47	0.140	.8229046 1.027903
pyperstudlep		1.000868	.0038063	0.23	0.820	.9934352 1.008356
pyperstude~s		.9865094	.0036949	-3.63	0.000	.9792942 .9937779
pyperstudw~e		.9955083	.0064886	-0.69	0.490	.9828717 1.008307
pyperstudb~k		.9951731	.0074085	-0.65	0.516	.9807582 1.0098
pyperstudl~o		1.001493	.0072743	0.21	0.837	.9873363 1.015852
pyperstudm~e		1.008143	.0059812	1.37	0.172	.9964883 1.019935
Asian		.6006242	.3229807	-0.95	0.343	.2093516 1.723175
Black		1.270802	.1503113	2.03	0.043	1.007853 1.602354
Latino		1.317673	.2113142	1.72	0.085	.9622793 1.804322
Other		.3335742	.1657402	-2.21	0.027	.1259688 .8833277
AsianMale		2.193723	1.722469	1.00	0.317	.4707986 10.22183
BlackMale		1.522148	.2988046	2.14	0.032	1.036006 2.236411
LatinoMale		1.769531	.4676376	2.16	0.031	1.054167 2.970346

OtherMale	5.785603	4.071751	2.49	0.013	1.456451	22.98271
-----						
3						
tenure	1.021909	.0065881	3.36	0.001	1.009078	1.034903
exper	1.08559	.0095938	9.29	0.000	1.066948	1.104557
male	1.129562	.1251002	1.10	0.271	.9091567	1.403401
age	.9626293	.0071585	-5.12	0.000	.9487006	.9767624
pyacctrati~2	1.336098	.0857996	4.51	0.000	1.178087	1.515304
pyperstudlep	1.020517	.0044316	4.68	0.000	1.011868	1.02924
pyperstude~s	1.011952	.0040692	2.95	0.003	1.004007	1.019959
pyperstudw~e	1.003536	.0070989	0.50	0.618	.9897188	1.017547
pyperstudb~k	.9943116	.0081164	-0.70	0.485	.9785303	1.010347
pyperstudl~o	.9780977	.0076208	-2.84	0.004	.9632746	.9931489
pyperstudm~e	.9924283	.007932	-0.95	0.342	.9770029	1.008097
Asian	1.491457	.7366657	0.81	0.418	.5664772	3.926804
Black	.8439365	.120623	-1.19	0.235	.6377469	1.116789
Latino	1.252386	.2201872	1.28	0.201	.8873299	1.767631
Other	.4915545	.2239019	-1.56	0.119	.2013035	1.200306
AsianMale	.3467197	.4129158	-0.89	0.374	.0335942	3.578429
BlackMale	1.933122	.4573611	2.79	0.005	1.215827	3.073596
LatinoMale	1.719197	.5117866	1.82	0.069	.9592504	3.081195
OtherMale	2.513778	1.983373	1.17	0.243	.5354613	11.80119
-----						

-----> region = 11

Iteration 0: log likelihood = -3214.5056  
Iteration 1: log likelihood = -3028.5521  
Iteration 2: log likelihood = -3024.6148  
Iteration 3: log likelihood = -3024.6058  
Iteration 4: log likelihood = -3024.6058

Multinomial logistic regression	Number of obs	=	2998
	LR chi2(38)	=	379.80
	Prob > chi2	=	0.0000
Log likelihood = -3024.6058	Pseudo R2	=	0.0591

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
tenure		1.024092	.0079073	3.08	0.002	1.008711	1.039708
exper		1.015806	.0089978	1.77	0.077	.9983226	1.033595
male		1.338009	.139865	2.79	0.005	1.090137	1.642242
age		.974065	.0067077	-3.82	0.000	.9610065	.9873009
pyacctrati~2		1.036614	.0701823	0.53	0.595	.9077949	1.183712
pyperstudlep		.9939229	.0044859	-1.35	0.177	.9851695	1.002754
pyperstude~s		.9934493	.0042354	-1.54	0.123	.9851827	1.001785
pyperstudw~e		.9972177	.0095284	-0.29	0.771	.9787161	1.016069
pyperstudb~k		.997612	.0111632	-0.21	0.831	.9759708	1.019733
pyperstudl~o		1.005757	.0105037	0.55	0.583	.9853791	1.026556
pyperstudm~e		1.01265	.0079289	1.61	0.108	.9972284	1.02831
Asian		2.04146	1.254603	1.16	0.246	.6121041	6.80858
Black		1.79955	.2989256	3.54	0.000	1.29948	2.492058
Latino		1.704021	.3430286	2.65	0.008	1.14848	2.528287
Other		.707667	.3063231	-0.80	0.424	.3029529	1.653038
AsianMale		1.797151	1.844553	0.57	0.568	.2403985	13.43499
BlackMale		1.451648	.4183564	1.29	0.196	.8251828	2.553716
LatinoMale		.9831966	.3249166	-0.05	0.959	.5144511	1.879042
OtherMale		1.04571	.7560092	0.06	0.951	.2535271	4.313185
3							
tenure		1.029379	.0079447	3.75	0.000	1.013925	1.045069
exper		1.085965	.0110543	8.10	0.000	1.064513	1.107848

male		1.509272	.1773348	3.50	0.000	1.19882	1.900121
age		.9686485	.0082761	-3.73	0.000	.9525627	.9850059
pyacctrati~2		1.485854	.1157124	5.08	0.000	1.275522	1.73087
pyperstudlep		1.012142	.0053622	2.28	0.023	1.001687	1.022707
pyperstude~s		1.021076	.004855	4.39	0.000	1.011605	1.030636
pyperstudw~e		1.019358	.0114607	1.71	0.088	.9971414	1.04207
pyperstudb~k		1.010218	.0133909	0.77	0.443	.9843105	1.036808
pyperstudl~o		.9997019	.0122499	-0.02	0.981	.9759786	1.024002
pyperstudm~e		.9937865	.0097854	-0.63	0.527	.9747913	1.013152
Asian		1.272455	1.002902	0.31	0.760	.2714921	5.963868
Black		1.548249	.320174	2.11	0.035	1.032321	2.322027
Latino		1.062057	.2650989	0.24	0.809	.6511493	1.732269
Other		.7714523	.3841233	-0.52	0.602	.2907213	2.047111
AsianMale		1.680841	2.182977	0.40	0.689	.1318367	21.42974
BlackMale		1.475513	.5306189	1.08	0.279	.7291839	2.985721
LatinoMale		1.832317	.7038	1.58	0.115	.8630798	3.890007
OtherMale		1.08336	.9071556	0.10	0.924	.2099015	5.591517

-----> region = 12

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -1001.2523  
Iteration 1: log likelihood = -937.23336  
Iteration 2: log likelihood = -935.75854  
Iteration 3: log likelihood = -935.62553  
Iteration 4: log likelihood = -935.60695  
Iteration 5: log likelihood = -935.6025  
Iteration 6: log likelihood = -935.60143  
Iteration 7: log likelihood = -935.60121  
Iteration 8: log likelihood = -935.60116  
Iteration 9: log likelihood = -935.60115

Multinomial logistic regression	Number of obs	=	931
	LR chi2(36)	=	131.30
	Prob > chi2	=	0.0000
Log likelihood = -935.60115	Pseudo R2	=	0.0656

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.011991	.0123692	0.98	0.329	.9880356 1.036527
exper		1.00947	.0156938	0.61	0.544	.9791747 1.040703
male		1.056979	.2026304	0.29	0.773	.7259133 1.539032
age		.9797165	.0120133	-1.67	0.095	.9564515 1.003547
pyacctrati~2		.8696858	.1067061	-1.14	0.255	.6837922 1.106116
pyperstudlep		1.012672	.0143556	0.89	0.374	.984923 1.041203
pyperstude~s		.9971604	.007534	-0.38	0.707	.9825028 1.012037
pyperstudw~e		.883992	.0427215	-2.55	0.011	.8041027 .9718185
pyperstudb~k		.89148	.0450021	-2.28	0.023	.8075004 .9841935
pyperstudl~o		.8953067	.0449973	-2.20	0.028	.8113182 .9879898
pyperstudm~e		1.018918	.0132427	1.44	0.149	.9932904 1.045206
Asian		4041212	7.91e+09	0.01	0.994	0 .
Black		.9758712	.3158361	-0.08	0.940	.5174938 1.840263
Latino		.3577916	.1662981	-2.21	0.027	.1438798 .8897347
Other		.5838574	.5226519	-0.60	0.548	.1010035 3.375027
AsianMale		(omitted)				
BlackMale		1.859745	1.001434	1.15	0.249	.6472904 5.343279
LatinoMale		3.121323	2.236721	1.59	0.112	.7662543 12.71465
OtherMale		7.48e-06	.0062518	-0.01	0.989	0 .
3						
tenure		1.005256	.0115227	0.46	0.647	.9829234 1.028095

exper		1.05815	.0177199	3.38	0.001	1.023983	1.093456
male		1.329643	.2566957	1.48	0.140	.9107626	1.941175
age		.9661353	.0135695	-2.45	0.014	.9399022	.9931005
pyacctrati~2		1.274792	.1603492	1.93	0.054	.9962565	1.6312
pyperstudlep		1.034216	.0148709	2.34	0.019	1.005477	1.063777
pyperstude~s		1.019097	.0080151	2.41	0.016	1.003508	1.034928
pyperstudw~e		1.002771	.0542037	0.05	0.959	.9019675	1.11484
pyperstudb~k		.9925815	.0564208	-0.13	0.896	.8879362	1.10956
pyperstudl~o		.9795234	.0552303	-0.37	0.714	.8770411	1.093981
pyperstudm~e		.9866889	.0137912	-0.96	0.338	.9600254	1.014093
Asian		.8762367	2769.318	-0.00	1.000	0	.
Black		1.197807	.4497374	0.48	0.631	.5738367	2.50026
Latino		.2776082	.1814208	-1.96	0.050	.0771188	.9993192
Other		.8607885	.772411	-0.17	0.867	.1482819	4.996948
AsianMale		(omitted)					
BlackMale		1.726868	1.013256	0.93	0.352	.5467751	5.453929
LatinoMale		5.376274	4.712781	1.92	0.055	.9645566	29.96643
OtherMale		4.648478	6.778616	1.05	0.292	.2667172	81.01595

-----> region = 13

```
Iteration 0: log likelihood = -1972.7792
Iteration 1: log likelihood = -1891.9657
Iteration 2: log likelihood = -1890.8018
Iteration 3: log likelihood = -1890.7305
Iteration 4: log likelihood = -1890.7216
Iteration 5: log likelihood = -1890.7204
Iteration 6: log likelihood = -1890.7202
Iteration 7: log likelihood = -1890.7202
Iteration 8: log likelihood = -1890.7202
```

Multinomial logistic regression	Number of obs	=	1823
	LR chi2(38)	=	164.12
	Prob > chi2	=	0.0000
Log likelihood = -1890.7202	Pseudo R2	=	0.0416

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.995991	.0097368	-0.41	0.681	.977089 1.015259
exper		.9851868	.0098983	-1.49	0.137	.9659762 1.004779
male		1.175896	.1652364	1.15	0.249	.8928092 1.548744
age		.9904755	.0082468	-1.15	0.250	.9744433 1.006772
pyacctrati~2		1.039777	.0930151	0.44	0.663	.8725583 1.239042
pyperstudlep		1.001818	.0058374	0.31	0.755	.9904416 1.013324
pyperstude~s		.9996807	.0059045	-0.05	0.957	.9881749 1.011132
pyperstudw~e		.9700734	.012664	-2.33	0.020	.9455673 .9952146
pyperstudb~k		.961449	.0161286	-2.34	0.019	.9303515 .993586
pyperstudl~o		.9771695	.0136092	-1.66	0.097	.9508567 1.00421
pyperstudm~e		1.003169	.0092439	0.34	0.731	.985214 1.021451
Asian		1.220875	1.014648	0.24	0.810	.239471 6.22429
Black		1.872027	.4545321	2.58	0.010	1.163158 3.012907
Latino		1.359432	.2570768	1.62	0.104	.9384044 1.969359
Other		.5237881	.3146194	-1.08	0.282	.1613881 1.699964
AsianMale		.407115	2.91e+08	0.02	0.986	0 .
BlackMale		.9970667	.4334507	-0.01	0.995	.4252897 2.337564
LatinoMale		1.140261	.3495945	0.43	0.669	.6252206 2.07958
OtherMale		.4695272	.5916551	-0.60	0.549	.0397241 5.549667
3						
tenure		1.024552	.0096493	2.58	0.010	1.005813 1.04364
exper		1.021115	.0111834	1.91	0.056	.9994292 1.043271

male		1.160897	.1770213	0.98	0.328	.8609877	1.565275
age		.9998577	.0094754	-0.02	0.988	.9814578	1.018603
pyacctrati~2		1.416312	.1386332	3.56	0.000	1.16907	1.715842
pyperstudlep		1.00673	.0064438	1.05	0.295	.9941794	1.019439
pyperstude~s		1.019818	.00636	3.15	0.002	1.007429	1.03236
pyperstudw~e		.9868233	.0144265	-0.91	0.364	.9589491	1.015508
pyperstudb~k		.9747794	.0183582	-1.36	0.175	.939454	1.011433
pyperstudl~o		.9769455	.0151157	-1.51	0.132	.9477641	1.007025
pyperstudm~e		.9854075	.0119272	-1.21	0.225	.9623058	1.009064
Asian		1.957807	1.526484	0.86	0.389	.4247132	9.024933
Black		1.264171	.3574161	0.83	0.407	.7263532	2.20021
Latino		.8221658	.1806843	-0.89	0.373	.5344339	1.264808
Other		.5140717	.3446085	-0.99	0.321	.138172	1.912615
AsianMale		335348.7	2.39e+08	0.02	0.986	0	.
BlackMale		.4384778	.2653633	-1.36	0.173	.1339082	1.435781
LatinoMale		1.501656	.5339064	1.14	0.253	.7480488	3.01447
OtherMale		1.522722	1.653819	0.39	0.699	.1811907	12.79691

-----> region = 14

note: Asian omitted because of collinearity  
note: AsianMale omitted because of collinearity  
note: BlackMale omitted because of collinearity  
note: OtherMale omitted because of collinearity  
Iteration 0: log likelihood = -276.05823  
Iteration 1: log likelihood = -235.67629  
Iteration 2: log likelihood = -232.86256  
Iteration 3: log likelihood = -232.61729  
Iteration 4: log likelihood = -232.59112  
Iteration 5: log likelihood = -232.58736  
Iteration 6: log likelihood = -232.5869  
Iteration 7: log likelihood = -232.5868  
Iteration 8: log likelihood = -232.58679

Multinomial logistic regression	Number of obs	=	277
	LR chi2(30)	=	86.94
	Prob > chi2	=	0.0000
Log likelihood = -232.58679	Pseudo R2	=	0.1575

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		.9738931	.0279875	-0.92	0.357	.9205549 1.030322
exper		1.058392	.0373603	1.61	0.108	.987643 1.134209
male		.8541537	.3728034	-0.36	0.718	.3630957 2.009329
age		1.003799	.0317395	0.12	0.905	.9434792 1.067976
pyacctrati~2		.5294515	.1690744	-1.99	0.046	.2831421 .9900289
pyperstudlep		.9841791	.0577312	-0.27	0.786	.8772901 1.104091
pyperstude~s		.9923763	.0206095	-0.37	0.713	.9527934 1.033604
pyperstudw~e		.8607486	.1408056	-0.92	0.359	.6246439 1.186097
pyperstudb~k		.8744119	.1679212	-0.70	0.485	.6001394 1.274031
pyperstudl~o		.8724478	.1432825	-0.83	0.406	.6323336 1.20374
pyperstudm~e		.8694485	.0578319	-2.10	0.035	.7631777 .9905172
Asian		(omitted)				
Black		1.85e+08	1.41e+12	0.00	0.998	0 .
Latino		8.43e-06	.0058061	-0.02	0.986	0 .
Other		1.707522	2266.375	0.00	1.000	0 .
AsianMale		(omitted)				
BlackMale		(omitted)				
LatinoMale		1420650	9.79e+08	0.02	0.984	0 .
OtherMale		(omitted)				

```

3      |
      tenure | .9906921 .0207966 -0.45 0.656 .9507588 1.032303
      exper | 1.044549 .0278172 1.64 0.102 .9914267 1.100518
      male | 2.799826 .9007051 3.20 0.001 1.490388 5.259724
      age | 1.026863 .0230704 1.18 0.238 .9826267 1.07309
pyacctrati~2 | 1.632993 .3569064 2.24 0.025 1.064011 2.50624
pyperstudlep | 1.01696 .0326095 0.52 0.600 .9550138 1.082925
pyperstude~s | 1.032373 .0129016 2.55 0.011 1.007393 1.057972
pyperstudw~e | 1.228277 .1376145 1.84 0.066 .9861177 1.529903
pyperstudb~k | 1.250777 .1568764 1.78 0.074 .9781813 1.599338
pyperstudl~o | 1.226963 .1374261 1.83 0.068 .9851273 1.528165
pyperstudm~e | .9778824 .0192362 -1.14 0.256 .9408977 1.016321
      Asian | (omitted)
      Black | .8215162 9117.685 -0.00 1.000 0 .
      Latino | .4537593 .728357 -0.49 0.623 .0195213 10.54731
      Other | 647706.3 4.23e+08 0.02 0.984 0 .
      AsianMale | (omitted)
      BlackMale | (omitted)
      LatinoMale | 12.07785 23.87138 1.26 0.207 .2509718 581.2387
      OtherMale | (omitted)
-----

```

```

-----> region = 15

```

```

note: Asian omitted because of collinearity
note: AsianMale omitted because of collinearity
note: BlackMale omitted because of collinearity
Iteration 0: log likelihood = -276.98708
Iteration 1: log likelihood = -249.06981
Iteration 2: log likelihood = -247.23182
Iteration 3: log likelihood = -247.01057
Iteration 4: log likelihood = -246.97461
Iteration 5: log likelihood = -246.97025
Iteration 6: log likelihood = -246.96963
Iteration 7: log likelihood = -246.96952
Iteration 8: log likelihood = -246.9695
Iteration 9: log likelihood = -246.96949

```

```

Multinomial logistic regression      Number of obs   =      265
                                      LR chi2(32)        =      60.04
                                      Prob > chi2        =      0.0019
Log likelihood = -246.96949          Pseudo R2       =      0.1084

```

```

-----
      admin |      RRR   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
1      |      (base outcome)
-----+-----
2      |
      tenure | 1.022545 .0340956    0.67  0.504   .9578555   1.091603
      exper | .9588885 .0351801   -1.14  0.253   .8923575   1.03038
      male | .6838338 .3350395   -0.78  0.438   .2617643   1.78645
      age | .9789441 .0283262   -0.74  0.462   .9249708   1.036067
pyacctrati~2 | .7065514 .2112285   -1.16  0.245   .3932522   1.269452
pyperstudlep | .9951424 .0345995   -0.14  0.889   .9295876   1.06532
pyperstude~s | 1.009128 .0172234    0.53  0.594   .9759295   1.043456
pyperstudw~e | 1.554078 .4192949    1.63  0.102   .9158329   2.637118
pyperstudb~k | 1.97895 .6369741    2.12  0.034   1.053063   3.718905
pyperstudl~o | 1.549716 .4176748    1.63  0.104   .9137745   2.628243
pyperstudm~e | .991106 .04817   -0.18  0.854   .9010518   1.090161
      Asian | (omitted)
      Black | 7.987503 10.71873    1.55  0.122   .575641   110.8333
      Latino | 3.071864 1.665572    2.07  0.038   1.061399   8.890484
      Other | 1.77e-06 .0024502   -0.01  0.992    0 .
      AsianMale | (omitted)
      BlackMale | (omitted)

```

LatinoMale		1.237571	1.078277	0.24	0.807	.2243524	6.826676
OtherMale		4380304	1.64e+10	0.00	0.997	0	.
-----							
3							
tenure		.9900908	.0216295	-0.46	0.648	.9485925	1.033405
exper		1.027059	.0282052	0.97	0.331	.9732395	1.083855
male		.958429	.324256	-0.13	0.900	.4938343	1.86011
age		.9607234	.0238546	-1.61	0.107	.9150887	1.008634
pyacctrati~2		1.08116	.2422549	0.35	0.728	.6968862	1.677329
pyperstudlep		1.050321	.028113	1.83	0.067	.9966412	1.106893
pyperstude~s		1.020342	.0133742	1.54	0.124	.9944627	1.046894
pyperstudw~e		1.270184	.2072539	1.47	0.143	.9225239	1.748863
pyperstudb~k		1.475964	.3144447	1.83	0.068	.9721487	2.240881
pyperstudl~o		1.235467	.2004862	1.30	0.193	.8988802	1.698089
pyperstudm~e		.9307485	.0326493	-2.05	0.041	.8689072	.9969911
Asian		(omitted)					
Black		8.87e-07	.0012798	-0.01	0.992	0	.
Latino		.9925539	.5126683	-0.01	0.988	.3606586	2.731567
Other		1.250456	1.808219	0.15	0.877	.0734835	21.27878
AsianMale		(omitted)					
BlackMale		(omitted)					
LatinoMale		1.642545	1.325343	0.62	0.539	.3378286	7.986159
OtherMale		2490948	4.77e+09	0.01	0.994	0	.
-----							

-----> region = 16

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -530.20525  
Iteration 1: log likelihood = -480.39618  
Iteration 2: log likelihood = -478.79823  
Iteration 3: log likelihood = -478.65732  
Iteration 4: log likelihood = -478.62877  
Iteration 5: log likelihood = -478.62268  
Iteration 6: log likelihood = -478.62133  
Iteration 7: log likelihood = -478.621  
Iteration 8: log likelihood = -478.62093  
Iteration 9: log likelihood = -478.62091

Multinomial logistic regression	Number of obs	=	488
	LR chi2(36)	=	103.17
	Prob > chi2	=	0.0000
Log likelihood = -478.62091	Pseudo R2	=	0.0973

admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1	(base outcome)					
2						
tenure	1.077125	.0222108	3.60	0.000	1.034461	1.121549
exper	.9123654	.0237285	-3.53	0.000	.8670239	.9600781
male	.9498477	.2626973	-0.19	0.852	.5523814	1.633311
age	1.005077	.0199873	0.25	0.799	.966656	1.045025
pyacctrati~2	1.289483	.2358416	1.39	0.165	.9010195	1.845429
pyperstudlep	1.025471	.0170184	1.52	0.130	.9926519	1.059375
pyperstude~s	.9939339	.0095929	-0.63	0.528	.9753088	1.012915
pyperstudw~e	.9970126	.0285687	-0.10	0.917	.9425623	1.054609
pyperstudb~k	1.032811	.0371632	0.90	0.370	.9624812	1.108279
pyperstudl~o	1.000442	.0270483	0.02	0.987	.9488086	1.054885
pyperstudm~e	1.020356	.0207598	0.99	0.322	.9804678	1.061866
Asian	1.17e-06	.0022997	-0.01	0.994	0	.
Black	.7898263	.7099175	-0.26	0.793	.1356589	4.598486
Latino	2.719417	1.78822	1.52	0.128	.7494578	9.867441
Other	8.46e-07	.0007265	-0.02	0.987	0	.
AsianMale	(omitted)					

BlackMale		381133.6	3.36e+08	0.01	0.988	0	.
LatinoMale		1.023882	.9111833	0.03	0.979	.1789532	5.858141
OtherMale		7574411	2.28e+10	0.01	0.996	0	.
-----							
3							
tenure		1.030824	.0162341	1.93	0.054	.9994918	1.063138
exper		.9875009	.0210007	-0.59	0.554	.9471863	1.029531
male		2.059966	.511508	2.91	0.004	1.266191	3.35136
age		1.004855	.0191206	0.25	0.799	.9680692	1.043038
pyacctrati~2		1.732756	.2909144	3.27	0.001	1.246891	2.407946
pyperstudlep		1.029687	.0158626	1.90	0.058	.9990613	1.061251
pyperstude~s		1.019115	.0092011	2.10	0.036	1.00124	1.037309
pyperstudw~e		1.038129	.0316099	1.23	0.219	.9779874	1.10197
pyperstudb~k		1.035565	.0396397	0.91	0.361	.9607154	1.116246
pyperstudl~o		1.025016	.0301758	0.84	0.401	.9675465	1.085899
pyperstudm~e		1.004389	.0214455	0.21	0.837	.9632242	1.047313
Asian		6.57e-07	.0011378	-0.01	0.993	0	.
Black		1.034702	1.017128	0.03	0.972	.1506856	7.104919
Latino		3.648905	2.369317	1.99	0.046	1.022017	13.02768
Other		2.738399	3.536057	0.78	0.435	.2179477	34.40657
AsianMale		(omitted)					
BlackMale		241966	2.13e+08	0.01	0.989	0	.
LatinoMale		.2172173	.2102926	-1.58	0.115	.0325708	1.44864
OtherMale		3891314	7.14e+09	0.01	0.993	0	.
-----							

-----> region = 17

note: AsianMale omitted because of collinearity

```

Iteration 0: log likelihood = -579.96423
Iteration 1: log likelihood = -506.91422
Iteration 2: log likelihood = -503.90663
Iteration 3: log likelihood = -503.67144
Iteration 4: log likelihood = -503.62758
Iteration 5: log likelihood = -503.61815
Iteration 6: log likelihood = -503.61598
Iteration 7: log likelihood = -503.61546
Iteration 8: log likelihood = -503.61536
Iteration 9: log likelihood = -503.61534

```

Multinomial logistic regression	Number of obs	=	554
	LR chi2(36)	=	152.70
	Prob > chi2	=	0.0000
Log likelihood = -503.61534	Pseudo R2	=	0.1316

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
tenure		1.021336	.0206781	1.04	0.297	.9816008 1.062679	
exper		.9864351	.0248016	-0.54	0.587	.9390031 1.036263	
male		2.025223	.533015	2.68	0.007	1.209056 3.392338	
age		.9825537	.0192199	-0.90	0.368	.9455963 1.020955	
pyacctrati~2		.9769487	.1733102	-0.13	0.895	.6900316 1.383167	
pyperstudlep		1.044126	.0252278	1.79	0.074	.9958325 1.094761	
pyperstude~s		.996435	.014614	-0.24	0.808	.9681998 1.025494	
pyperstudw~e		.9463045	.0708839	-0.74	0.461	.8170917 1.095951	
pyperstudb~k		.9588624	.0759752	-0.53	0.596	.8209404 1.119956	
pyperstudl~o		.9363165	.0729637	-0.84	0.398	.8036958 1.090821	
pyperstudm~e		1.001278	.0183169	0.07	0.944	.9660138 1.03783	
Asian		8.02e+07	2.41e+11	0.01	0.995	0	
Black		1.627426	1.316077	0.60	0.547	.3335389 7.94065	
Latino		.432562	.2839481	-1.28	0.202	.1194789 1.566049	
Other		1.198813	1.521434	0.14	0.886	.0996501 14.42199	



AsianMale		(omitted)					
BlackMale		1229054	1.26e+09	0.01	0.989	0	.
LatinoMale		3.28938	2.634407	1.49	0.137	.6845441	15.80617
OtherMale		4.27e-07	.0007502	-0.01	0.993	0	.
-----							
3							
tenure		1.045012	.0179611	2.56	0.010	1.010395	1.080815
exper		1.037122	.0234529	1.61	0.107	.9921588	1.084123
male		4.010696	1.038712	5.36	0.000	2.414188	6.662981
age		.9861301	.0187108	-0.74	0.462	.9501311	1.023493
pyacctrati~2		1.156293	.2023461	0.83	0.407	.8205622	1.629388
pyperstudlep		1.013771	.0209457	0.66	0.508	.9735383	1.055666
pyperstude~s		1.063545	.0144085	4.55	0.000	1.035676	1.092163
pyperstudw~e		1.080991	.094644	0.89	0.374	.9105355	1.283357
pyperstudb~k		1.023807	.0936382	0.26	0.797	.8557888	1.224813
pyperstudl~o		1.047527	.0934671	0.52	0.603	.8794587	1.247714
pyperstudm~e		.9392098	.0220745	-2.67	0.008	.896926	.9834871
Asian		3.441649	15559.53	0.00	1.000	0	.
Black		3.178804	3.130219	1.17	0.240	.4613941	21.90057
Latino		1.325517	.6227467	0.60	0.549	.5278114	3.328832
Other		9.34e-07	.0010567	-0.01	0.990	0	.
AsianMale		(omitted)					
BlackMale		1838113	1.89e+09	0.01	0.989	0	.
LatinoMale		.5545808	.3894776	-0.84	0.401	.1400145	2.196629
OtherMale		964788.7	1.09e+09	0.01	0.990	0	.
-----							

-----> region = 18

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -414.76237  
Iteration 1: log likelihood = -378.26022  
Iteration 2: log likelihood = -376.95452  
Iteration 3: log likelihood = -376.76888  
Iteration 4: log likelihood = -376.7355  
Iteration 5: log likelihood = -376.72788  
Iteration 6: log likelihood = -376.72616  
Iteration 7: log likelihood = -376.72575  
Iteration 8: log likelihood = -376.72566  
Iteration 9: log likelihood = -376.72565

Multinomial logistic regression	Number of obs	=	380
	LR chi2(36)	=	76.07
	Prob > chi2	=	0.0001
Log likelihood = -376.72565	Pseudo R2	=	0.0917

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	1.016448	.0220009	0.75	0.451	.9742287	1.060497
	exper	.9729487	.0265718	-1.00	0.315	.9222383	1.026448
	male	1.068523	.3852317	0.18	0.854	.5271118	2.166034
	age	.9895112	.0210557	-0.50	0.620	.9490916	1.031652
pyacctrati~2		.7621315	.1721473	-1.20	0.229	.4895126	1.186577
pyperstudlep		.9834727	.0191707	-0.85	0.393	.9466074	1.021774
pyperstude~s		.9992092	.0135158	-0.06	0.953	.9730667	1.026054
pyperstudw~e		1.020868	.139805	0.15	0.880	.7805489	1.335178
pyperstudb~k		1.155318	.1716238	0.97	0.331	.8634844	1.545783
pyperstudl~o		1.023053	.1402686	0.17	0.868	.7819727	1.338457
pyperstudm~e		1.0391	.031072	1.28	0.200	.9799503	1.10182
	Asian	8.45e+07	2.99e+11	0.01	0.996	0	.
	Black	5.221499	6.214565	1.39	0.165	.506647	53.81272
	Latino	2.49869	1.048269	2.18	0.029	1.09802	5.686104

Other		.6169591	2821.132	-0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		.5873499	.9005334	-0.35	0.729	.0290953	11.85691
LatinoMale		.8522971	.5331289	-0.26	0.798	.2501194	2.904254
OtherMale		3.4785	22494.38	0.00	1.000	0	.
-----							
3							
tenure		1.037096	.0200564	1.88	0.060	.9985215	1.07716
exper		1.014521	.0273439	0.53	0.593	.9623192	1.069555
male		1.204938	.3874313	0.58	0.562	.641612	2.262856
age		.9675075	.0205219	-1.56	0.119	.9281099	1.008578
pyacctrati~2		1.054864	.2198922	0.26	0.798	.7010661	1.587209
pyperstudlep		.9976565	.0158259	-0.15	0.882	.9671156	1.029162
pyperstude~s		1.023268	.0130655	1.80	0.072	.9979777	1.049199
pyperstudw~e		.9432966	.1095253	-0.50	0.615	.7513045	1.184351
pyperstudb~k		.9588075	.1229819	-0.33	0.743	.7456786	1.232853
pyperstudl~o		.9251984	.1091075	-0.66	0.510	.7342665	1.165778
pyperstudm~e		.9692353	.0266512	-1.14	0.256	.9183825	1.022904
Asian		1.224555	6374.239	0.00	1.000	0	.
Black		4.543761	5.436143	1.27	0.206	.4355524	47.40134
Latino		1.14407	.4567467	0.34	0.736	.5231514	2.501944
Other		6009618	1.80e+10	0.01	0.996	0	.
AsianMale		(omitted)					
BlackMale		1.57e-07	.000141	-0.02	0.986	0	.
LatinoMale		1.170748	.6907365	0.27	0.789	.3683475	3.721078
OtherMale		1.668032	7052.28	0.00	1.000	0	.
-----							

-----> region = 19

note: AsianMale omitted because of collinearity

Iteration 0: log likelihood = -881.4415  
Iteration 1: log likelihood = -810.07665  
Iteration 2: log likelihood = -806.72665  
Iteration 3: log likelihood = -806.30206  
Iteration 4: log likelihood = -806.22779  
Iteration 5: log likelihood = -806.22054  
Iteration 6: log likelihood = -806.21894  
Iteration 7: log likelihood = -806.21858  
Iteration 8: log likelihood = -806.21852

Multinomial logistic regression	Number of obs	=	814
	LR chi2(36)	=	150.45
	Prob > chi2	=	0.0000
Log likelihood = -806.21852	Pseudo R2	=	0.0853

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
-----						
1						
tenure		.993856	.0146741	-0.42	0.676	.9655074 1.023037
exper		.9529956	.0175928	-2.61	0.009	.9191307 .9881082
male		1.002313	.3422702	0.01	0.995	.5132569 1.957366
age		1.016662	.0135133	1.24	0.214	.990518 1.043495
pyacctrati~2		1.183741	.1731532	1.15	0.249	.8886816 1.576766
pyperstudlep		1.002963	.0062217	0.48	0.633	.9908425 1.015232
pyperstude~s		1.013283	.009772	1.37	0.171	.99431 1.032618
pyperstudw~e		1.107431	.1035226	1.09	0.275	.9220324 1.330108
pyperstudb~k		1.01878	.0904882	0.21	0.834	.8560051 1.212507
pyperstudl~o		1.061972	.0876055	0.73	0.466	.90343 1.248335
pyperstudm~e		1.004777	.0102677	0.47	0.641	.9848533 1.025104
Asian		2887864	5.31e+09	0.01	0.994	0
Black		.9851742	.4361261	-0.03	0.973	.4137092 2.346016
Latino		.6461371	.1588675	-1.78	0.076	.3990593 1.046193
Other		.9383853	.8918718	-0.07	0.947	.1456705 6.044922
AsianMale		(omitted)				

BlackMale		.848751	.6677066	-0.21	0.835	.1816126	3.966566
LatinoMale		.8950654	.3620912	-0.27	0.784	.4050497	1.977886
OtherMale		.5918133	.8331688	-0.37	0.709	.0374837	9.343865
-----							
2		(base outcome)					
-----							
3							
tenure		.9803203	.0142693	-1.37	0.172	.9527481	1.00869
exper		1.118881	.0240731	5.22	0.000	1.07268	1.167073
male		1.22997	.4699344	0.54	0.588	.5816696	2.600836
age		.9609172	.0171852	-2.23	0.026	.9278184	.9951967
pyacctrati~2		1.271923	.212675	1.44	0.150	.916502	1.765176
pyperstudlep		1.025531	.0069596	3.71	0.000	1.011981	1.039263
pyperstude~s		1.01118	.0115309	0.97	0.330	.9888303	1.034034
pyperstudw~e		1.013799	.0917643	0.15	0.880	.8489947	1.210594
pyperstudb~k		1.008363	.0876211	0.10	0.924	.8504572	1.195588
pyperstudl~o		.9852522	.077462	-0.19	0.850	.8445487	1.149397
pyperstudm~e		.9796847	.0162443	-1.24	0.216	.9483583	1.012046
Asian		.9006143	2621.288	-0.00	1.000	0	.
Black		.3911419	.2504287	-1.47	0.143	.1115199	1.371881
Latino		.6807644	.1855155	-1.41	0.158	.3990574	1.161337
Other		2.32e-06	.0016638	-0.02	0.986	0	.
AsianMale		(omitted)					
BlackMale		3.12e-06	.0013803	-0.03	0.977	0	.
LatinoMale		.9543648	.4318375	-0.10	0.918	.3931488	2.316711
OtherMale		640066.2	4.59e+08	0.02	0.985	0	.

-----> region = 20

Iteration 0: log likelihood = -2192.6506  
Iteration 1: log likelihood = -2092.8672  
Iteration 2: log likelihood = -2089.876  
Iteration 3: log likelihood = -2089.576  
Iteration 4: log likelihood = -2089.5225  
Iteration 5: log likelihood = -2089.5108  
Iteration 6: log likelihood = -2089.5081  
Iteration 7: log likelihood = -2089.5074  
Iteration 8: log likelihood = -2089.5073  
Iteration 9: log likelihood = -2089.5073

Multinomial logistic regression	Number of obs	=	2078
	LR chi2(38)	=	206.29
	Prob > chi2	=	0.0000
Log likelihood = -2089.5073	Pseudo R2	=	0.0470

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2							
	tenure	.9912517	.0082409	-1.06	0.291	.9752307	1.007536
	exper	1.010277	.0100087	1.03	0.302	.9908491	1.030085
	male	1.356364	.2056402	2.01	0.044	1.007685	1.825695
	age	.9832278	.0077823	-2.14	0.033	.9680925	.9985998
pyacctrati~2		.9406998	.077418	-0.74	0.458	.8005687	1.105359
pyperstudlep		.9987321	.0066589	-0.19	0.849	.9857659	1.011869
pyperstude~s		.9975373	.0057004	-0.43	0.666	.986427	1.008773
pyperstudw~e		1.004125	.0274779	0.15	0.880	.9516881	1.059451
pyperstudb~k		1.011036	.0309662	0.36	0.720	.9521295	1.073588
pyperstudl~o		1.006118	.0286558	0.21	0.830	.9514921	1.063879
pyperstudm~e		.9991628	.0082218	-0.10	0.919	.9831776	1.015408
	Asian	.6335529	.5231698	-0.55	0.580	.1255698	3.196544
	Black	.78198	.234443	-0.82	0.412	.4345098	1.407316
	Latino	1.379032	.2004326	2.21	0.027	1.03719	1.833541

Other		.1588035	.1667316	-1.75	0.080	.0202844	1.243246
AsianMale		4776419	5.13e+09	0.01	0.989	0	.
BlackMale		2.104511	.9905399	1.58	0.114	.8365893	5.294076
LatinoMale		1.07009	.2424236	0.30	0.765	.6864119	1.66823
OtherMale		6.837845	8.572665	1.53	0.125	.5858152	79.81379
-----							
3							
tenure		1.014877	.0086455	1.73	0.083	.9980723	1.031964
exper		1.059369	.0124948	4.89	0.000	1.03516	1.084143
male		1.244105	.2139231	1.27	0.204	.8881647	1.742692
age		.975075	.0097728	-2.52	0.012	.9561074	.9944188
pyacctrati~2		1.135383	.1026781	1.40	0.160	.9509641	1.355566
pyperstudlep		1.023743	.0067992	3.53	0.000	1.010503	1.037156
pyperstude~s		1.018751	.0062184	3.04	0.002	1.006636	1.031012
pyperstudw~e		1.047045	.0319035	1.51	0.131	.9863457	1.111479
pyperstudb~k		1.02648	.0350013	0.77	0.443	.9601212	1.097426
pyperstudl~o		1.022908	.0322972	0.72	0.473	.9615252	1.088209
pyperstudm~e		1.006639	.0088457	0.75	0.451	.9894507	1.024127
Asian		.3273322	.3600873	-1.02	0.310	.0378975	2.827269
Black		1.393781	.3994135	1.16	0.247	.7948172	2.444117
Latino		.9865194	.1610928	-0.08	0.934	.7163239	1.358632
Other		.4961113	.3936475	-0.88	0.377	.1047562	2.349518
AsianMale		1.29e+07	1.39e+10	0.02	0.988	0	.
BlackMale		1.124889	.5862874	0.23	0.821	.4050109	3.124297
LatinoMale		.9564434	.2568019	-0.17	0.868	.5650859	1.618841
OtherMale		1.81e-06	.0012744	-0.02	0.985	0	.
-----							

by pyschooltype:

```
. bysort pyschooltype : mlogit admin tenure exper male age pyacctrating2 pyperstudlep
pyperstuddecodis pyperstudwhite pyperstudblack pyperstudlatino pyperstudmobile Asian
Black Latino
> Other AsianMale BlackMale LatinoMale OtherMale if validcert2011a==1, rrr iter(20)
```

```
-----> pyschooltype =
no observations
```

```
-----> pyschooltype = B
```

```
note: AsianMale omitted because of collinearity
Iteration 0: log likelihood = -422.57399
Iteration 1: log likelihood = -392.87974
Iteration 2: log likelihood = -387.05933
Iteration 3: log likelihood = -386.34912
Iteration 4: log likelihood = -386.29743
Iteration 5: log likelihood = -386.28696
Iteration 6: log likelihood = -386.28537
Iteration 7: log likelihood = -386.28519
Iteration 8: log likelihood = -386.28515
Iteration 9: log likelihood = -386.28514
```

```
Multinomial logistic regression      Number of obs   =      464
                                     LR chi2(36)       =      72.58
                                     Prob > chi2      =      0.0003
Log likelihood = -386.28514          Pseudo R2       =      0.0859
```

	admin	RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
1		(base outcome)					
2	tenure	1.032523	.0319835	1.03	0.302	.9717013	1.097151

exper		.9678165	.0324519	-0.98	0.329	.9062569	1.033558
male		.7152772	.2963862	-0.81	0.419	.3175162	1.611324
age		.9750218	.0256739	-0.96	0.337	.9259783	1.026663
pyacctrati~2		.8042206	.200197	-0.88	0.381	.4937227	1.309988
pyperstudlep		.9735139	.0272274	-0.96	0.337	.9215855	1.028368
pyperstude~s		.9712554	.0125899	-2.25	0.024	.9468905	.9962472
pyperstudw~e		.9529758	.0276618	-1.66	0.097	.900273	1.008764
pyperstudb~k		.9971881	.0327593	-0.09	0.932	.9350045	1.063507
pyperstudl~o		.9846861	.030634	-0.50	0.620	.9264386	1.046596
pyperstudm~e		1.013338	.020872	0.64	0.520	.9732446	1.055084
Asian		4.248731	7.672155	0.80	0.423	.1233673	146.3249
Black		1.531056	1.452888	0.45	0.654	.2383679	9.834097
Latino		1.46511	1.056009	0.53	0.596	.3567374	6.017165
Other		3.988943	13838.52	0.00	1.000	0	.
AsianMale		(omitted)					
BlackMale		1.177478	1.837744	0.10	0.917	.0552641	25.08779
LatinoMale		2.980951	3.030739	1.07	0.283	.4063837	21.8662
OtherMale		.7134162	2994.679	-0.00	1.000	0	.
-----							
3							
tenure		.9521112	.0163932	-2.85	0.004	.9205173	.9847895
exper		1.021354	.017771	1.21	0.225	.9871109	1.056785
male		1.151505	.2661557	0.61	0.542	.7320173	1.811382
age		.9986452	.015417	-0.09	0.930	.968881	1.029324
pyacctrati~2		1.070248	.1602319	0.45	0.650	.7980815	1.435232
pyperstudlep		1.006934	.0156044	0.45	0.656	.9768098	1.037988
pyperstude~s		.9987234	.0086345	-0.15	0.883	.9819426	1.015791
pyperstudw~e		.9716842	.0228254	-1.22	0.221	.9279615	1.017467
pyperstudb~k		.9835989	.0257889	-0.63	0.528	.9343304	1.035465
pyperstudl~o		.9762015	.0241831	-0.97	0.331	.9299359	1.024769
pyperstudm~e		1.00585	.0133485	0.44	0.660	.9800247	1.032356
Asian		8.10e-08	.0002941	-0.00	0.996	0	.
Black		.7431394	.6817532	-0.32	0.746	.1230782	4.487034
Latino		.8554174	.5314148	-0.25	0.802	.2531525	2.890506
Other		2822344	3.70e+09	0.01	0.991	0	.
AsianMale		(omitted)					
BlackMale		2.431877	3.020142	0.72	0.474	.2132244	27.73617
LatinoMale		3.875507	3.188208	1.65	0.100	.7728279	19.43454
OtherMale		1.351479	2170.427	0.00	1.000	0	.
-----							

-----> pyschooltype = E

Iteration 0: log likelihood = -12631.536  
Iteration 1: log likelihood = -12005.791  
Iteration 2: log likelihood = -12000.805  
Iteration 3: log likelihood = -12000.803  
Iteration 4: log likelihood = -12000.803

Multinomial logistic regression	Number of obs	=	11602
	LR chi2(38)	=	1261.47
	Prob > chi2	=	0.0000
Log likelihood = -12000.803	Pseudo R2	=	0.0499

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.01526	.0042498	3.62	0.000	1.006965 1.023624
exper		1.015888	.0052155	3.07	0.002	1.005717 1.026162
male		1.486982	.1249385	4.72	0.000	1.261208 1.753174
age		.9765138	.0037278	-6.23	0.000	.9692347 .9838476
pyacctrati~2		.9311817	.0319879	-2.08	0.038	.8705505 .9960357
pyperstudlep		1.00367	.001687	2.18	0.029	1.000369 1.006982

pyperstude~s		.9915395	.0020363	-4.14	0.000	.9875564	.9955386
pyperstudw~e		1.004946	.0044155	1.12	0.261	.9963293	1.013638
pyperstudb~k		1.003031	.0051533	0.59	0.556	.9929813	1.013182
pyperstudl~o		1.007539	.004664	1.62	0.105	.9984393	1.016722
pyperstudm~e		1.011822	.0039887	2.98	0.003	1.004035	1.01967
Asian		.6870643	.2224428	-1.16	0.246	.3642619	1.295928
Black		1.005221	.090042	0.06	0.954	.8433655	1.19814
Latino		1.15706	.0822469	2.05	0.040	1.006585	1.330031
Other		.7920841	.1974389	-0.94	0.350	.4859554	1.291059
AsianMale		2.786311	1.97895	1.44	0.149	.6925832	11.20953
BlackMale		1.047294	.2024714	0.24	0.811	.7169807	1.529783
LatinoMale		.8030651	.1119836	-1.57	0.116	.6110192	1.055472
OtherMale		.8723577	.5489856	-0.22	0.828	.2541106	2.99479

3							
tenure		1.019099	.0037353	5.16	0.000	1.011804	1.026446
exper		1.074687	.0052709	14.69	0.000	1.064406	1.085068
male		2.220355	.1708488	10.37	0.000	1.909525	2.581781
age		.9815451	.0038401	-4.76	0.000	.9740475	.9891005
pyacctrati~2		.9741666	.0330226	-0.77	0.440	.9115468	1.041088
pyperstudlep		.9951662	.0016602	-2.90	0.004	.9919176	.9984253
pyperstude~s		1.0041	.0019878	2.07	0.039	1.000211	1.008003
pyperstudw~e		1.009001	.0045182	2.00	0.045	1.000184	1.017895
pyperstudb~k		.9980233	.0051888	-0.38	0.704	.9879051	1.008245
pyperstudl~o		1.004972	.0047315	1.05	0.292	.995741	1.014288
pyperstudm~e		1.004482	.0039818	1.13	0.259	.996708	1.012317
Asian		.697027	.2426974	-1.04	0.300	.3522657	1.379205
Black		.9737491	.088524	-0.29	0.770	.8148243	1.163671
Latino		.9620742	.0695181	-0.54	0.593	.8350297	1.108448
Other		.9031867	.2195865	-0.42	0.675	.5608283	1.454538
AsianMale		2.367753	1.729709	1.18	0.238	.5656013	9.912028
BlackMale		.6619629	.1324714	-2.06	0.039	.4471899	.9798855
LatinoMale		.6217931	.085976	-3.44	0.001	.4741866	.815347
OtherMale		.6426003	.3836546	-0.74	0.459	.1994087	2.070797

-----> pyschooltype = M

Iteration 0: log likelihood = -6875.679  
Iteration 1: log likelihood = -6559.8647  
Iteration 2: log likelihood = -6552.8223  
Iteration 3: log likelihood = -6552.7994  
Iteration 4: log likelihood = -6552.7994

Multinomial logistic regression	Number of obs	=	6517
	LR chi2(38)	=	645.76
	Prob > chi2	=	0.0000
Log likelihood = -6552.7994	Pseudo R2	=	0.0470

admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]
1		(base outcome)				
2						
tenure		1.009917	.0049558	2.01	0.044	1.000251 1.019677
exper		1.016292	.0058904	2.79	0.005	1.004812 1.027903
male		1.67972	.1261163	6.91	0.000	1.449864 1.946017
age		.9814513	.0044584	-4.12	0.000	.9727517 .9902286
pyacctrati~2		.8680311	.0428253	-2.87	0.004	.7880254 .9561594
pyperstudlep		1.00283	.0034175	0.83	0.407	.9961544 1.009551
pyperstude~s		.9967166	.0026352	-1.24	0.214	.991565 1.001895
pyperstudw~e		.9892777	.0059265	-1.80	0.072	.9777299 1.000962
pyperstudb~k		.9879707	.006933	-1.72	0.085	.9744753 1.001653
pyperstudl~o		.9906815	.0064323	-1.44	0.149	.9781543 1.003369
pyperstudm~e		1.007732	.0045584	1.70	0.089	.9988375 1.016706

Asian		1.423581	.5811603	0.87	0.387	.6395714	3.168657
Black		1.196802	.1223069	1.76	0.079	.9795671	1.462214
Latino		1.87495	.2056578	5.73	0.000	1.512251	2.324639
Other		.5185199	.1795717	-1.90	0.058	.2630152	1.022233
AsianMale		.6255095	.4075329	-0.72	0.471	.1744447	2.2429
BlackMale		1.109552	.1845329	0.63	0.532	.8009067	1.53714
LatinoMale		.911849	.1401671	-0.60	0.548	.6746498	1.232445
OtherMale		2.547936	1.295646	1.84	0.066	.9404714	6.902896
-----							
3							
tenure		1.013925	.0052751	2.66	0.008	1.003639	1.024317
exper		1.089482	.0079902	11.69	0.000	1.073933	1.105255
male		2.149763	.1829308	8.99	0.000	1.819529	2.539933
age		.9525635	.0060287	-7.68	0.000	.9408205	.9644531
pyacctrati~2		.8210836	.0480353	-3.37	0.001	.7321333	.920841
pyperstudlep		.9897337	.0043373	-2.35	0.019	.9812691	.9982714
pyperstude~s		1.011061	.0031431	3.54	0.000	1.00492	1.01724
pyperstudw~e		.9999603	.0077637	-0.01	0.996	.9848589	1.015293
pyperstudb~k		.9827312	.0089189	-1.92	0.055	.9654051	1.000368
pyperstudl~o		.9888614	.0082672	-1.34	0.180	.9727901	1.005198
pyperstudm~e		1.000211	.0057925	0.04	0.971	.9889218	1.011629
Asian		1.666168	.7992673	1.06	0.287	.6507199	4.266222
Black		.8701703	.1212627	-1.00	0.318	.6621936	1.143467
Latino		1.617872	.2204822	3.53	0.000	1.238634	2.113222
Other		.7570533	.3028286	-0.70	0.487	.3456509	1.658117
AsianMale		.3268749	.302489	-1.21	0.227	.0532945	2.004844
BlackMale		1.450975	.2990514	1.81	0.071	.9687776	2.17318
LatinoMale		.8758654	.1602777	-0.72	0.469	.6118896	1.253723
OtherMale		.7966042	.5146063	-0.35	0.725	.2245776	2.825653
-----							

-----> pyschooltype = S

Iteration 0: log likelihood = -7594.3094  
Iteration 1: log likelihood = -7263.8269  
Iteration 2: log likelihood = -7253.3052  
Iteration 3: log likelihood = -7253.0962  
Iteration 4: log likelihood = -7253.0496  
Iteration 5: log likelihood = -7253.0381  
Iteration 6: log likelihood = -7253.0358  
Iteration 7: log likelihood = -7253.0355  
Iteration 8: log likelihood = -7253.0354  
Iteration 9: log likelihood = -7253.0354

Multinomial logistic regression	Number of obs	=	7854
	LR chi2(38)	=	682.55
	Prob > chi2	=	0.0000
Log likelihood = -7253.0354	Pseudo R2	=	0.0449

-----							
admin		RRR	Std. Err.	z	P> z	[95% Conf. Interval]	
-----							
1		(base outcome)					
-----							
2							
tenure		1.009863	.0038796	2.55	0.011	1.002288	1.017495
exper		1.012365	.004794	2.60	0.009	1.003012	1.021805
male		1.215762	.0775215	3.06	0.002	1.072934	1.377604
age		.9745013	.0038982	-6.46	0.000	.9668909	.9821716
pyacctrati~2		.9439258	.0409143	-1.33	0.183	.8670471	1.027621
pyperstudlep		1.020906	.0046563	4.54	0.000	1.01182	1.030073
pyperstude~s		.9884174	.0022619	-5.09	0.000	.9839942	.9928605
pyperstudw~e		.9915862	.0054472	-1.54	0.124	.980967	1.00232
pyperstudb~k		.9960592	.0062986	-0.62	0.532	.9837903	1.008481
pyperstudl~o		.9964106	.0058984	-0.61	0.544	.9849169	1.008039
pyperstudm~e		1.004413	.0041918	1.06	0.291	.9962305	1.012662

	Asian		1.029841	.3989531	0.08	0.939	.4819708	2.200491
	Black		1.518412	.1557297	4.07	0.000	1.241909	1.856478
	Latino		1.862044	.1937103	5.98	0.000	1.518583	2.283186
	Other		.5717091	.218359	-1.46	0.143	.2704373	1.208603
	AsianMale		2.344809	1.54938	1.29	0.197	.6421827	8.561626
	BlackMale		1.554299	.2281519	3.00	0.003	1.165705	2.072433
	LatinoMale		1.185369	.1599723	1.26	0.208	.9098692	1.544287
	OtherMale		1.794956	.9357928	1.12	0.262	.6460757	4.986825
-----								
3								
	tenure		.9983981	.0047691	-0.34	0.737	.9890945	1.007789
	exper		1.073552	.0077602	9.82	0.000	1.05845	1.08887
	male		2.085091	.1877987	8.16	0.000	1.74767	2.487657
	age		.9581079	.0063963	-6.41	0.000	.9456531	.9707268
	pyacctrati~2		1.086659	.0673629	1.34	0.180	.9623359	1.227044
	pyperstudlep		1.001273	.0072029	0.18	0.860	.9872548	1.015491
	pyperstude~s		1.014261	.0031437	4.57	0.000	1.008119	1.020442
	pyperstudw~e		1.02355	.0099644	2.39	0.017	1.004205	1.043268
	pyperstudb~k		1.012028	.0111177	1.09	0.276	.9904702	1.034054
	pyperstudl~o		1.010742	.0104073	1.04	0.299	.9905489	1.031347
	pyperstudm~e		.9880941	.0069847	-1.69	0.090	.9744988	1.001879
	Asian		1.312712	.831181	0.43	0.667	.3794956	4.540796
	Black		1.148911	.212116	0.75	0.452	.8000818	1.649826
	Latino		1.525268	.2567223	2.51	0.012	1.096675	2.12136
	Other		.4939821	.3636051	-0.96	0.338	.116726	2.090522
	AsianMale		1.97e-06	.000993	-0.03	0.979	0	.
	BlackMale		1.139	.2760657	0.54	0.591	.7082937	1.831615
	LatinoMale		.8879515	.1810333	-0.58	0.560	.5954543	1.324128
	OtherMale		4.23932	3.494246	1.75	0.080	.8427446	21.32536
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